

# A Simple and Efficient Error-Diffusion Algorithm

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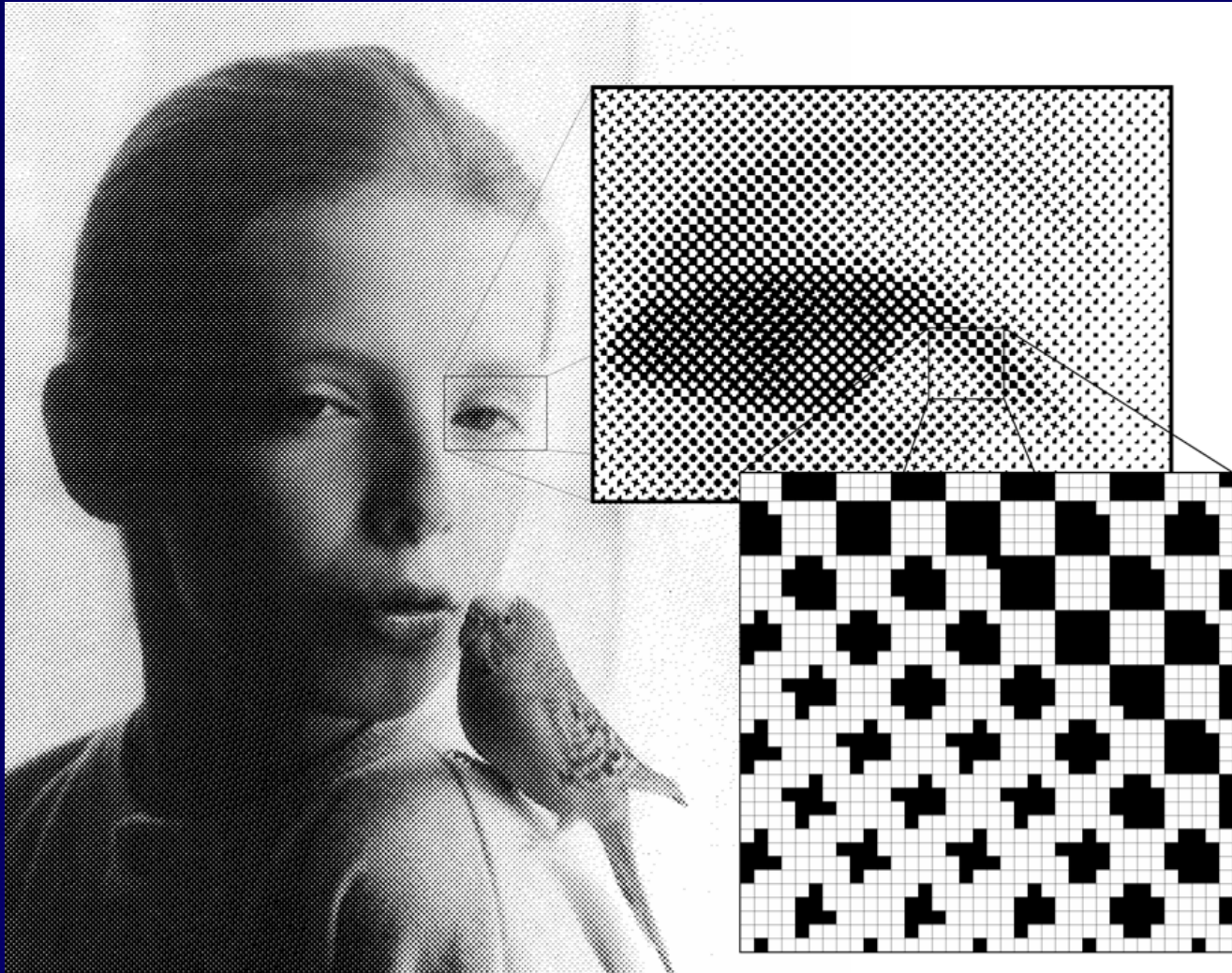
## Motivations

- **Error-Diffusion is Important Visualization**
- **No Satisfactory Solution Exists Today**
- **A Simple and Efficient Solution is Possible**

# Outline

- **Introduction**
- **Problem Statement**
- **Our Variable-Coefficient Error-Diffusion Algorithm**
- **Results**
- **Conclusions and Future Work**

# What is Halftoning?





# Where Do We Need Halftoning?

Displays



PDAs, Mobile Phones



Games

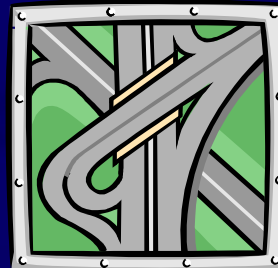


Target Devices for Our Algorithm: Low Resolution, Where Individual Pixels Are Visible

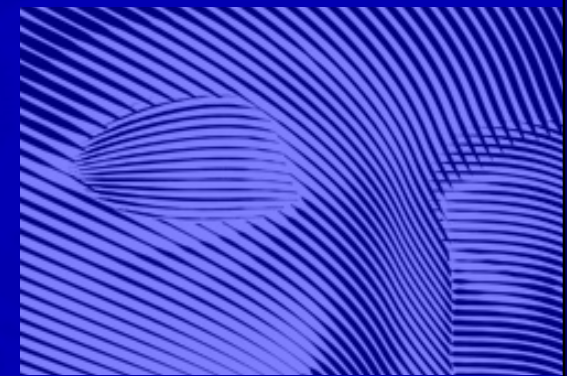
Printers



Network-Based Imaging



Artistic Rendering

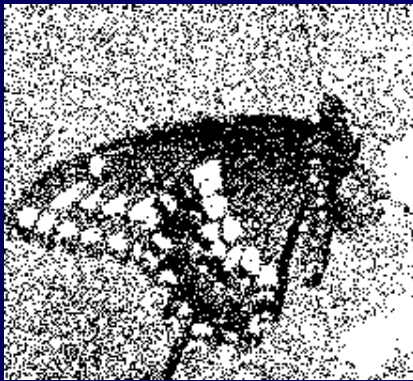


# Halftoning: Basic Classification

Continuous-Tone Original



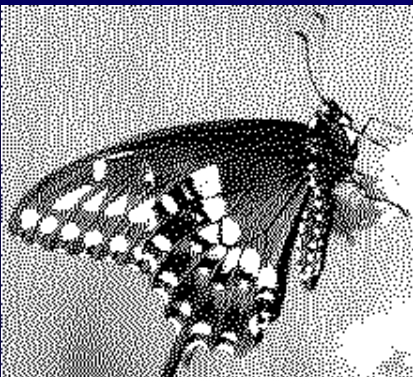
Monte-Carlo



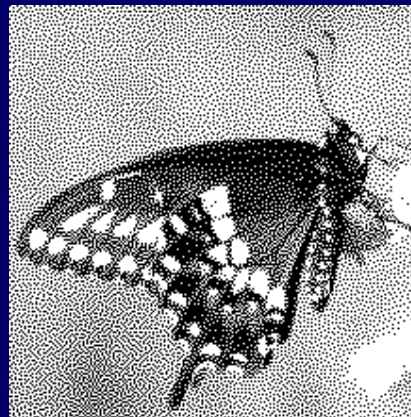
Clustered-Dot Dither



Error-Diffusion



DBS-based



Blue Noise Mask



# Halftoning Algorithms: General Requirements

- **Visual Quality**
- **Execution Speed**
- **Conceptual Simplicity**
- **Legal Availability**



# Criterion: Visual Quality for Low-Resolution Displays

Continuous-Tone Original



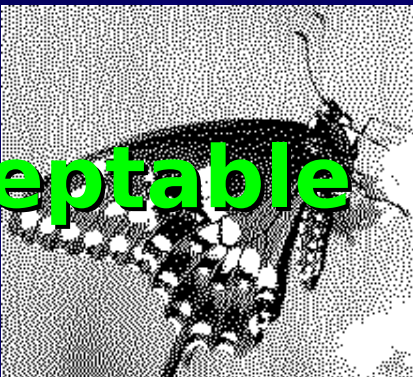
Monte-Carlo



Clustered-Dot Dither



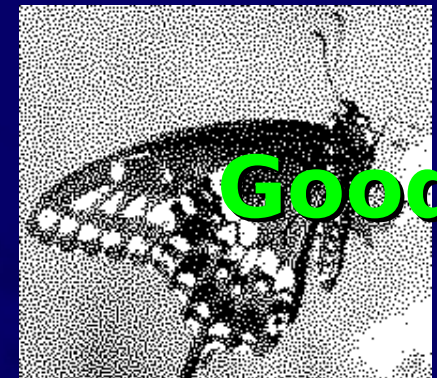
Floyd-Steinberg E-D



DBS-based



Blue Noise Mask



Acceptable

Excellent

Good

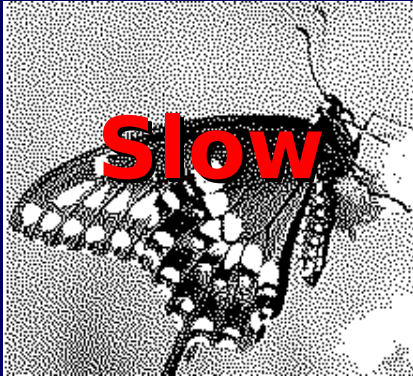


## Criterion: Execution Speed

Continuous-Tone Original



Marcu's Roadmap



Clustered-Dot Dither



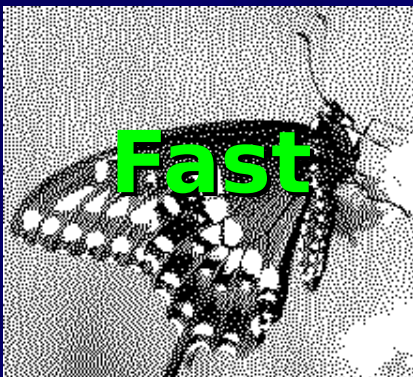
**Fast**

Blue Noise Mask



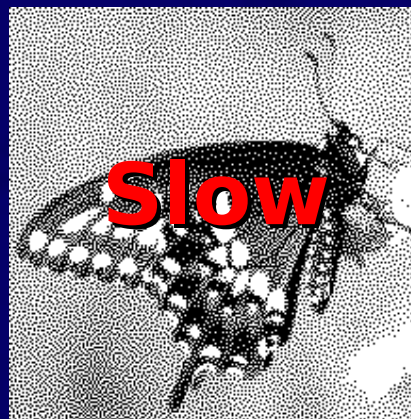
**SIGGRAPH**  
2001 EXPLORE INTERACTION  
AND DIGITAL IMAGERY

Floyd-Steinberg E-D



**Fast**

DBS-based



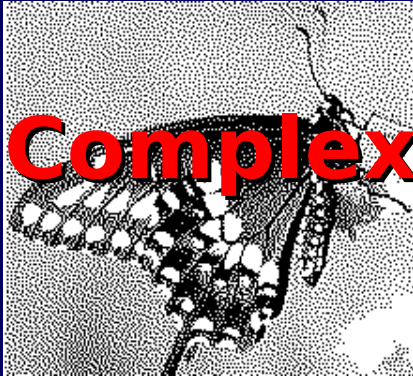
**Slow**

## Criterion: Conceptual Simplicity

Continuous-Tone Original



Marcu's Roadmap



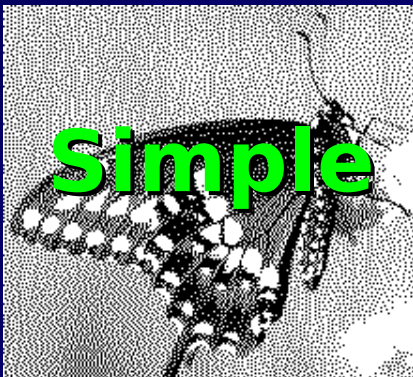
**Complex**

Clustered-Dot Dither



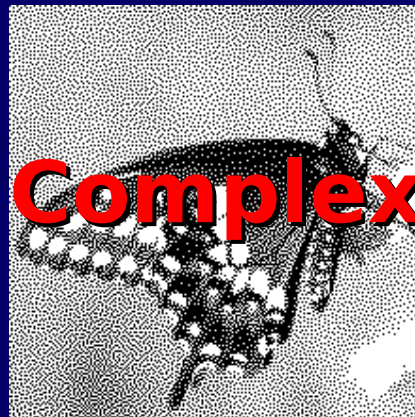
**Simple**  
Blue Noise Mask

Error-Diffusion



**Simple**

DBS-based

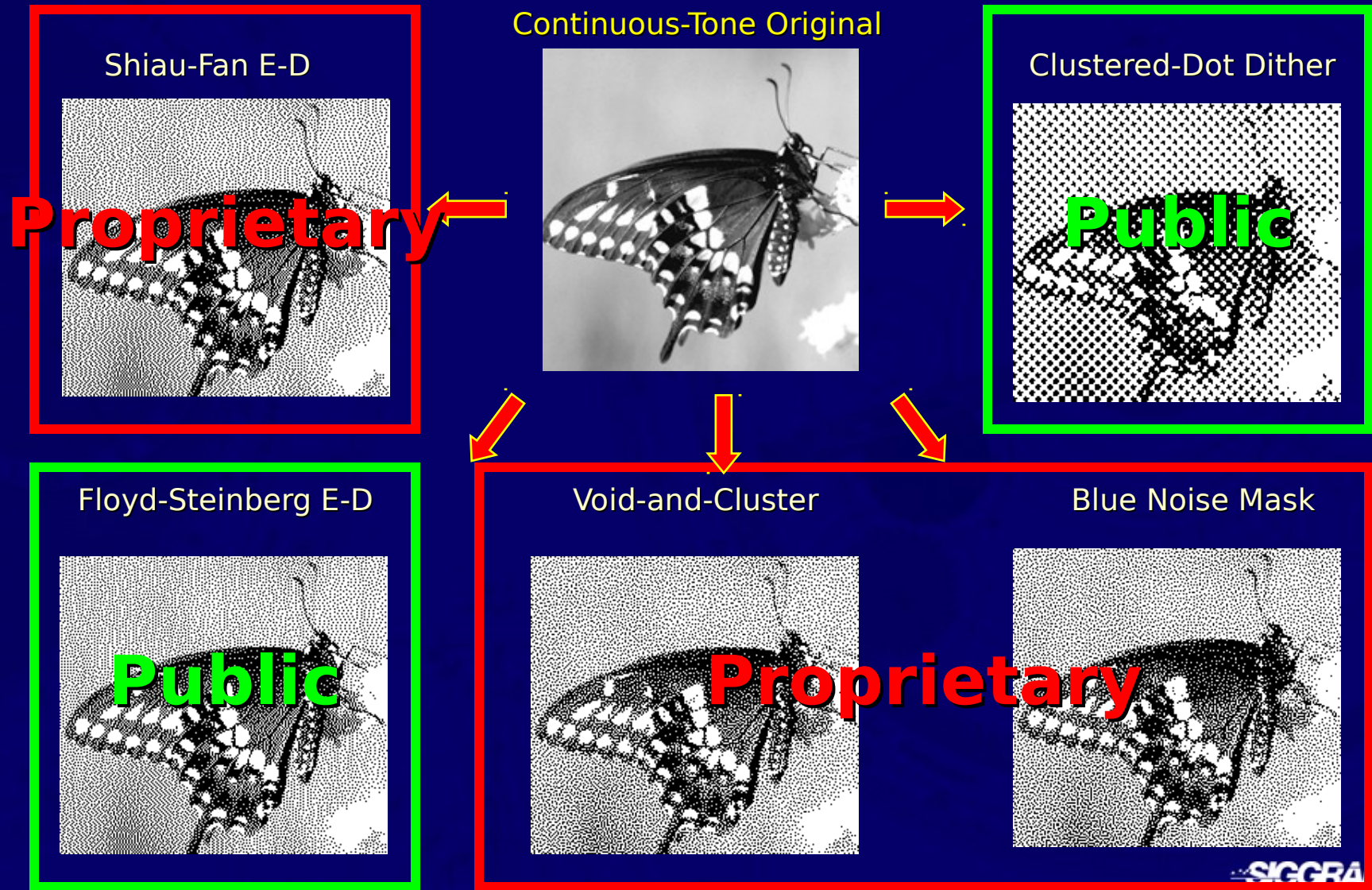


**Complex**





## Criterion: Public Availability



## State-of-the-Art Status

**There is **NO** Halftoning Algorithm  
that Wold Fully Satisfy All  
Criteria**

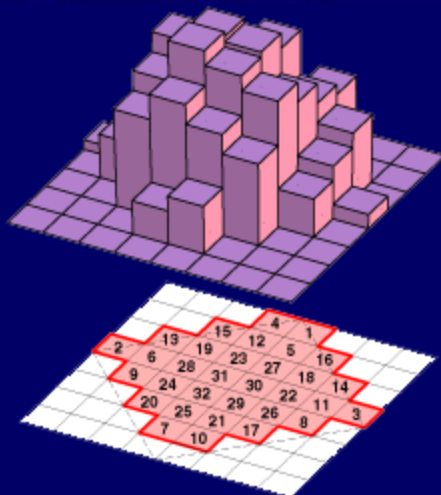
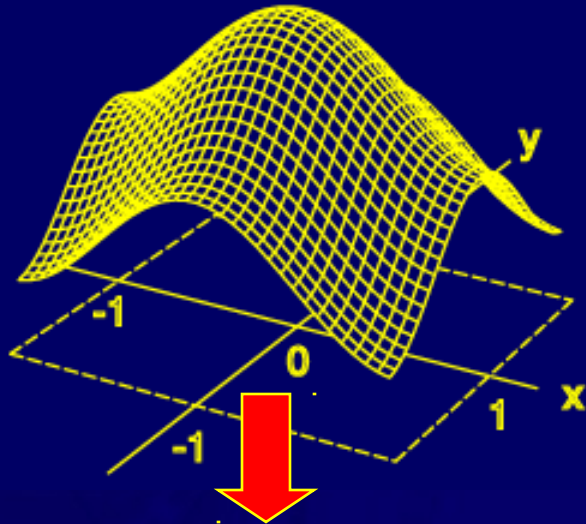
- **Visual Quality**
- **Execution Speed**
- **Conceptual Simplicity**
- **Legal Availability**

**→ Main Motivation for Our  
Work**



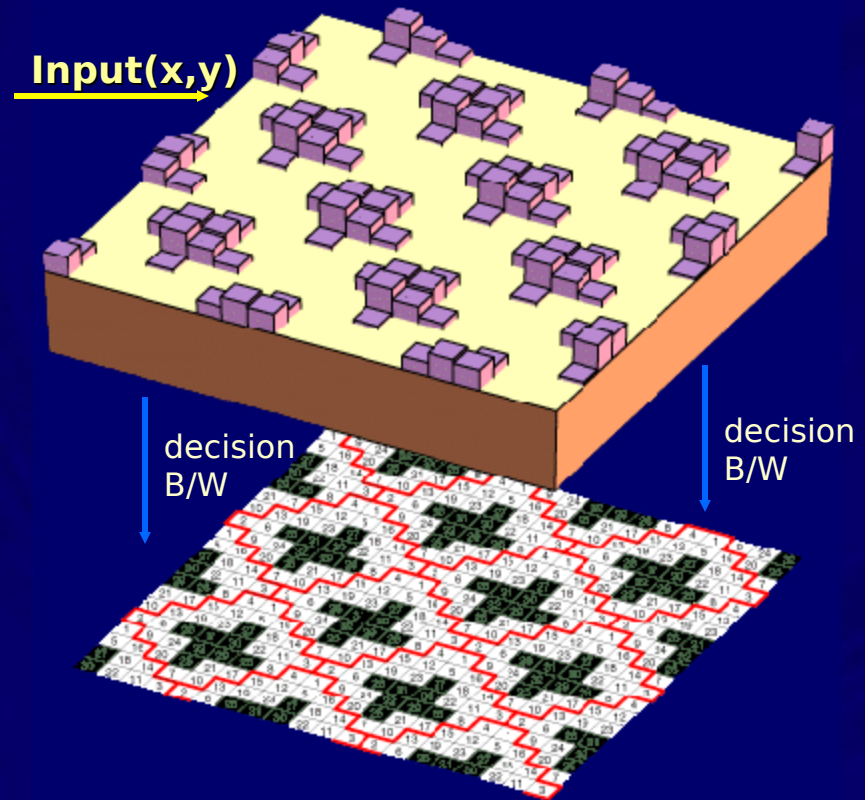
# Clustered-Dot Dithering

Spot Function (here: Egg-crate function)

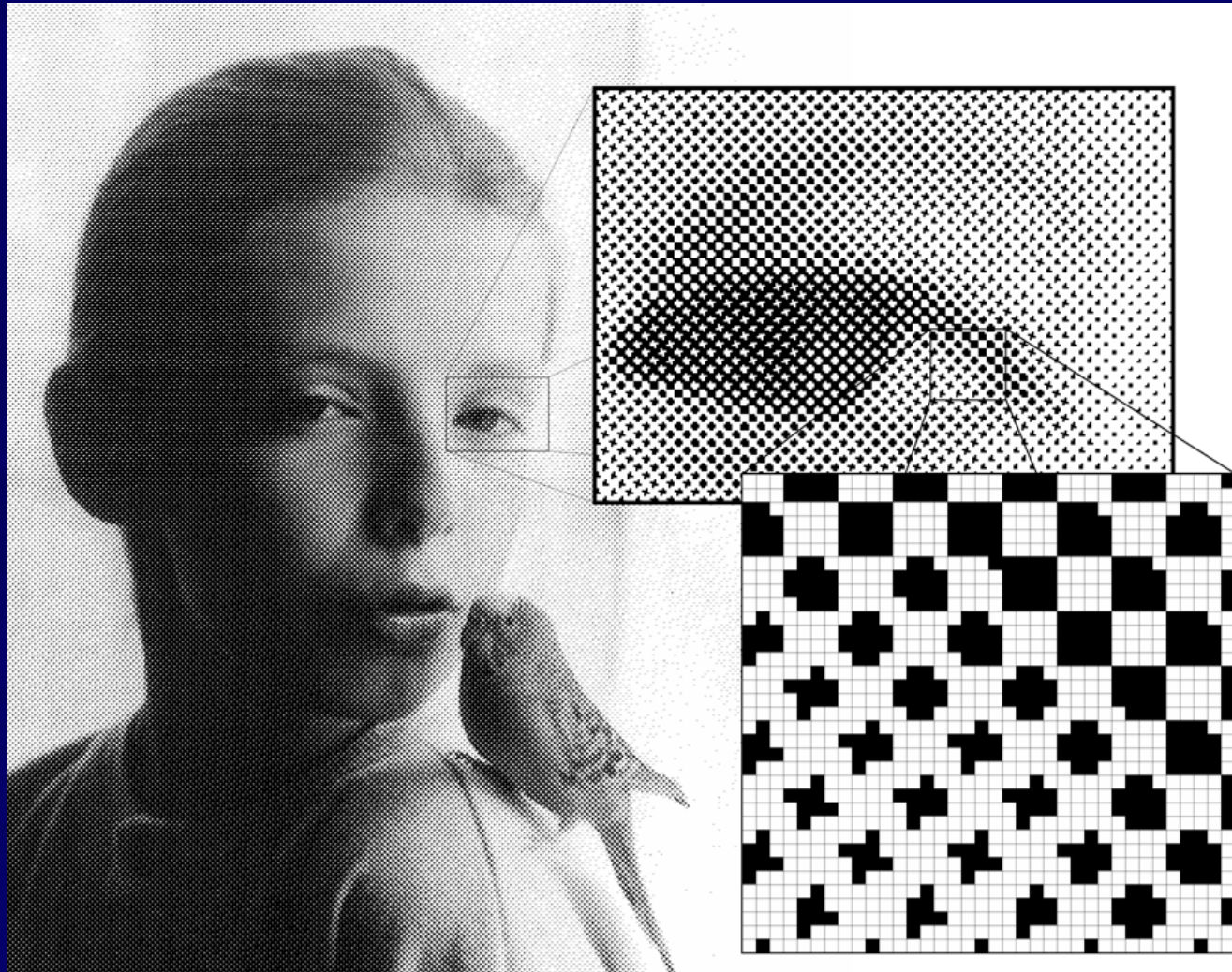


**Threshold(x,y)**

Compare **Threshold(x,y)** and **Input(x,y)**

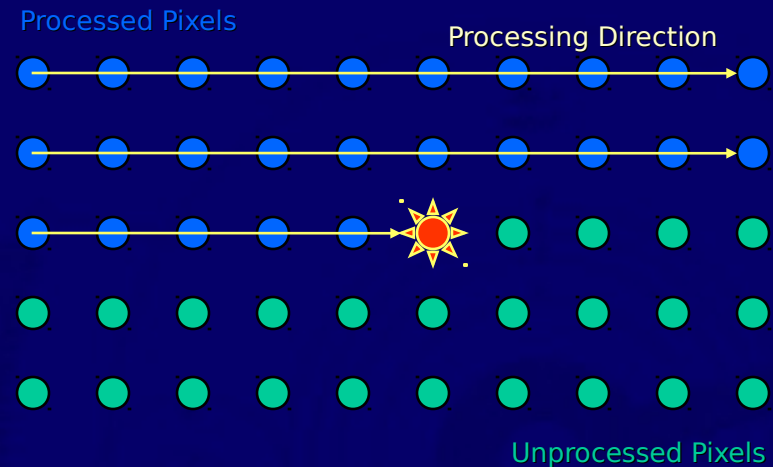


# Clustered-Dot Dithering



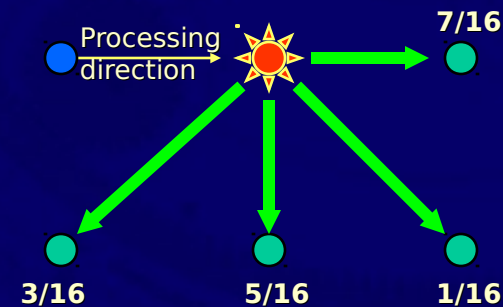
# Floyd-Steinberg Error-Diffusion Algorithm (1975)

- **Process Pixels in Order of Scanlines**



- **Compare  $\text{Input}(x,y)$  with  $\text{Threshold}(x,y)=.5$**

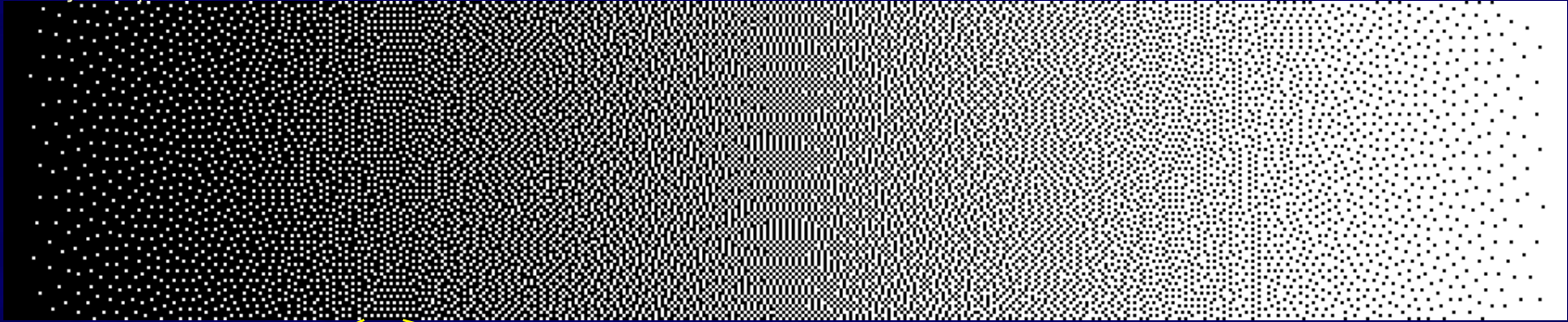
- **Distribute Quantization Error on Unprocessed Pixels**



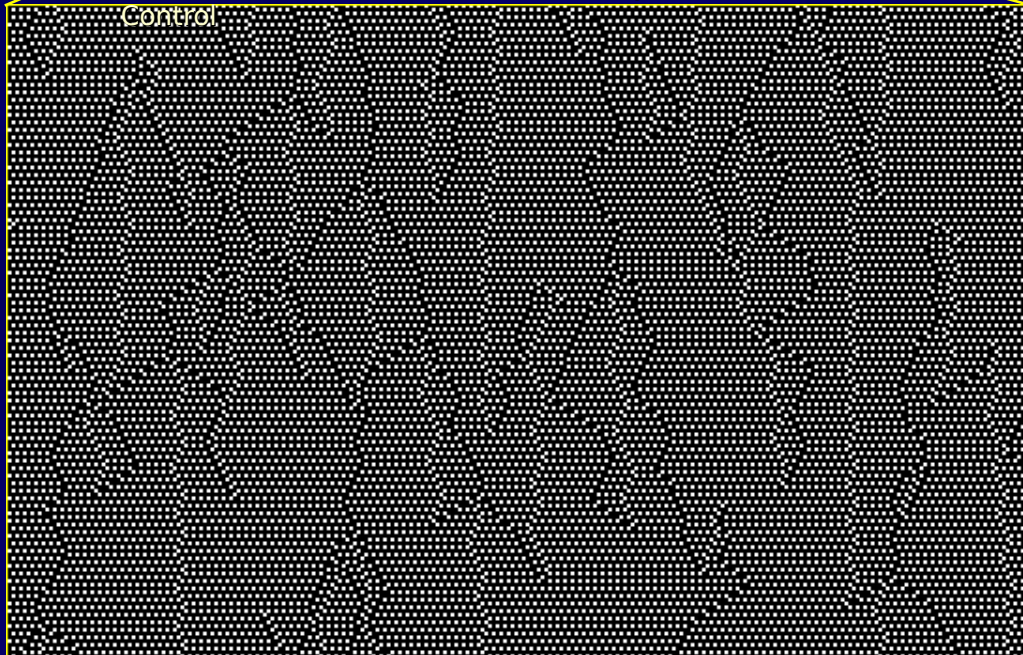


# Floyd-Steinberg Error-Diffusion Algorithm

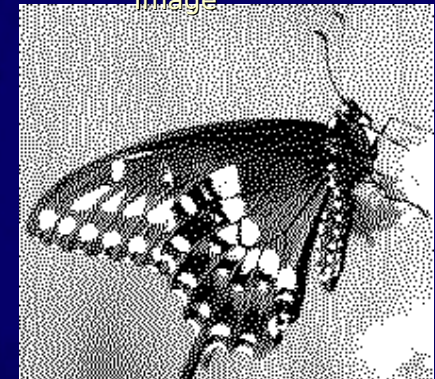
Gray Ramp



Intolerable Artifacts at Certain Intensity Levels, Out of Control.



Sample Image

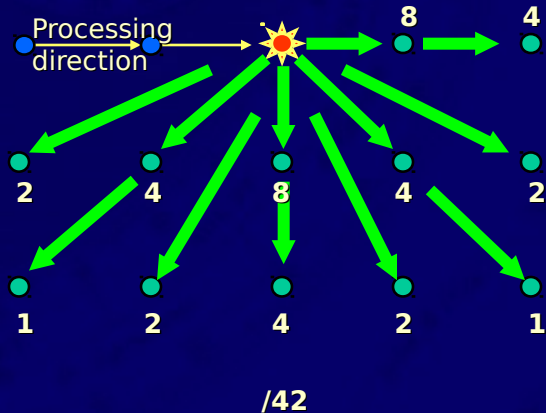




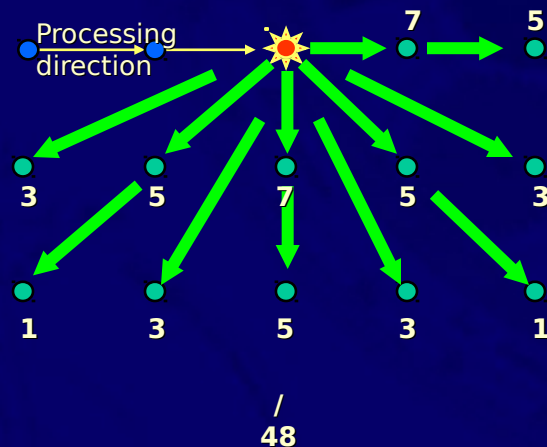
# Sophistication of Error-Diffusion Algorithms

- **Processing Paths: Serpentine, Space-Filling Curves**
- **Different Error Distribution Coefficients**
- **Threshold Modulation: Control of Edge Enhancement**

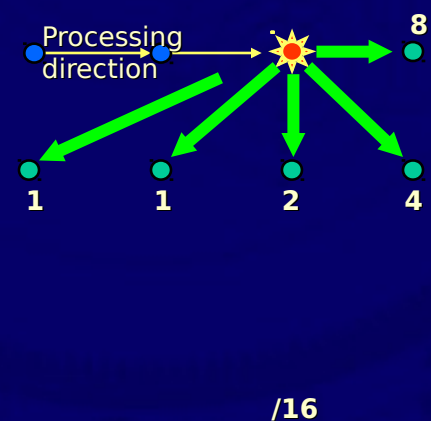
Stucki (1981)



Judice-Jarvice-Ninke (1976)

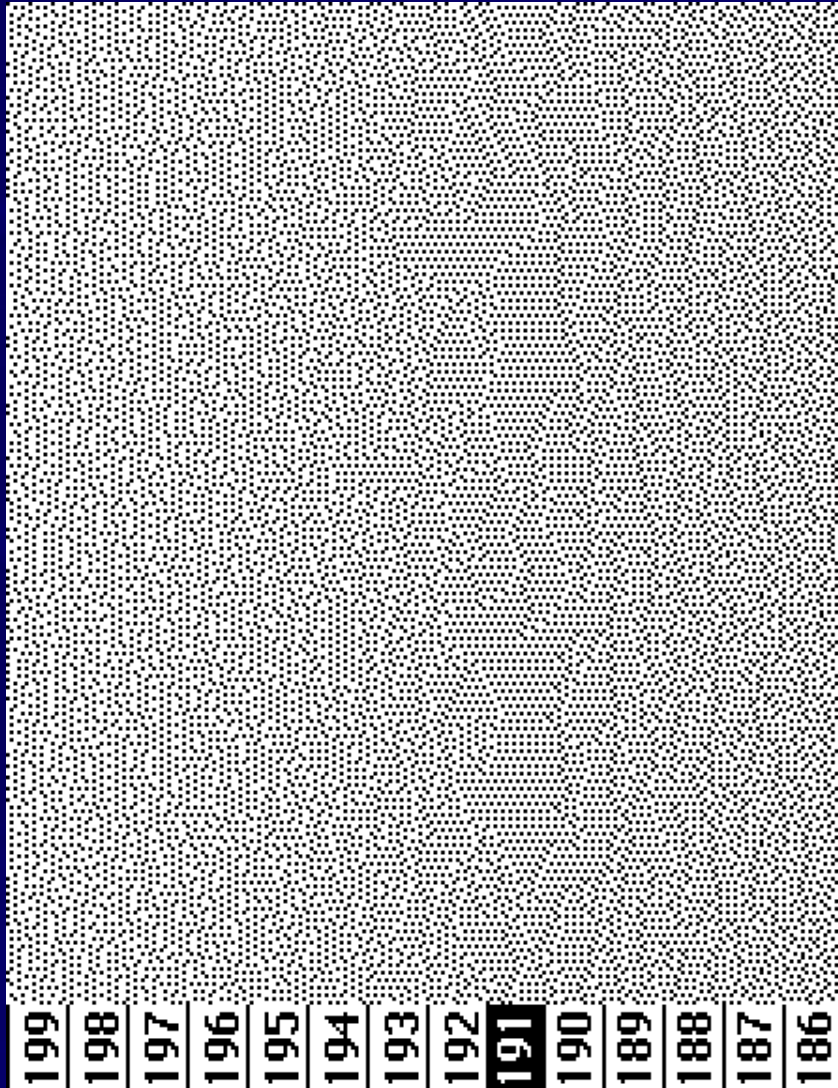


Shiau-Fan (1996)

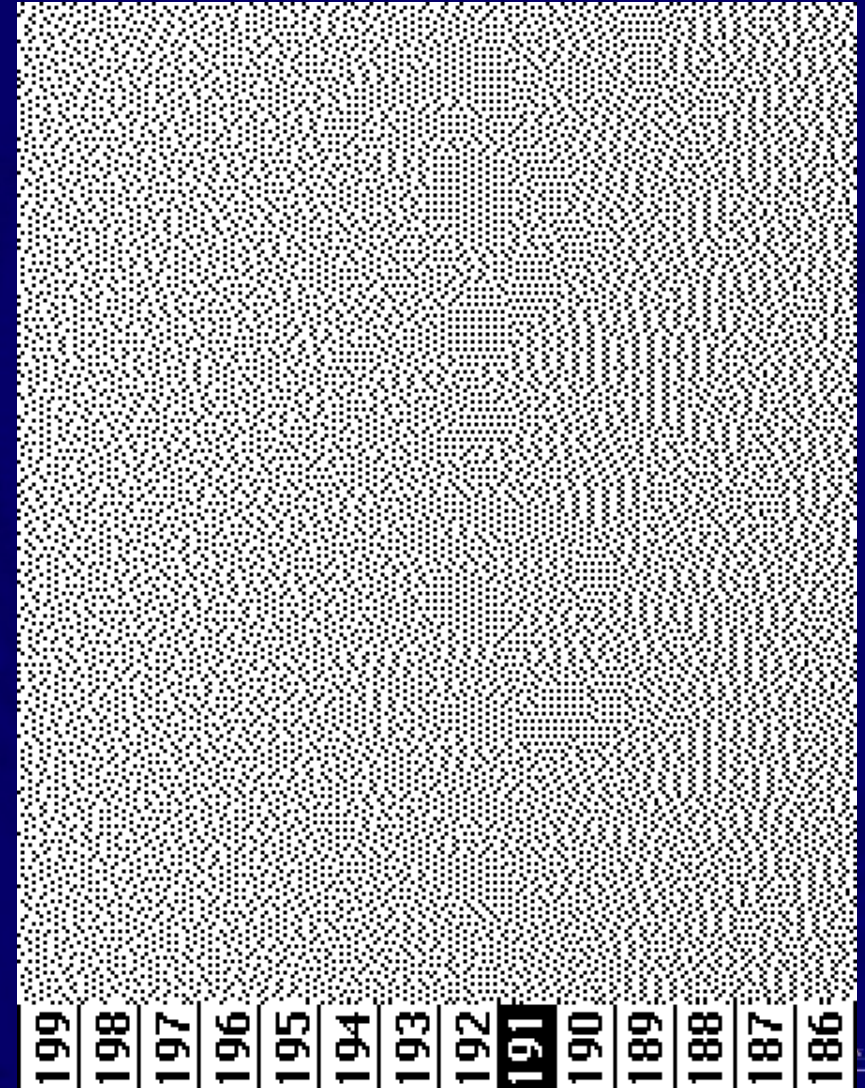


# Hidden Artifacts

Std. Floyd-Steinberg E-D



Shiau-Fan E-D



# Different Error-Diffusion Algorithms

Floyd-Steinberg



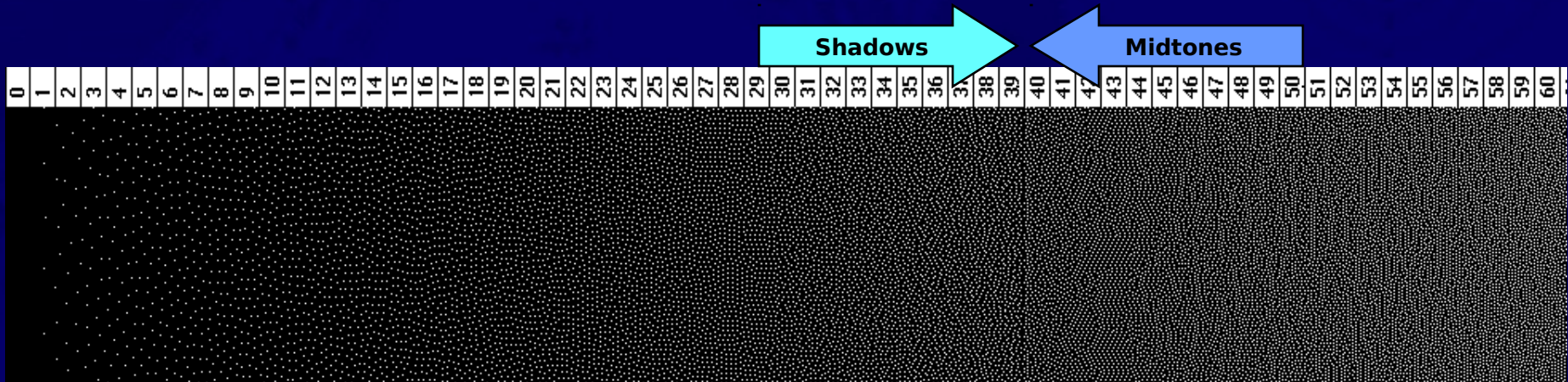
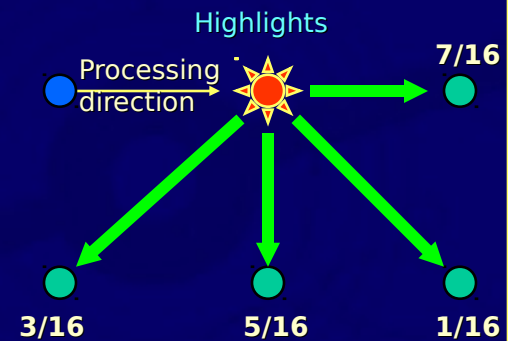
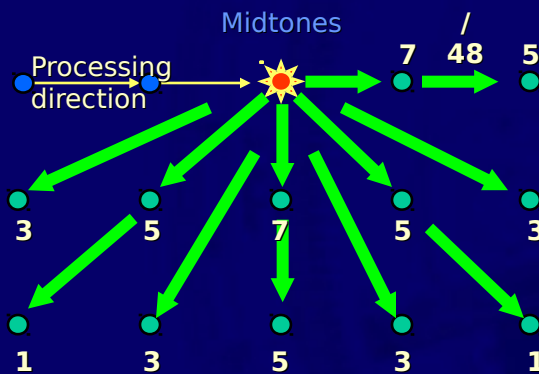
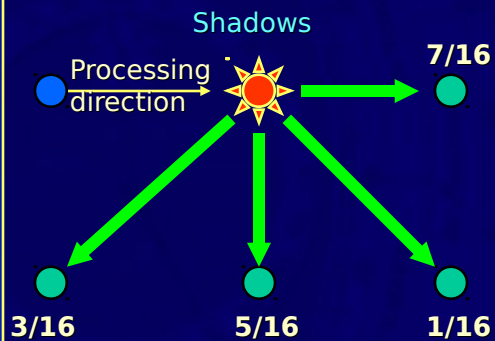
Stucki



Judice-Jarvis-Ninke

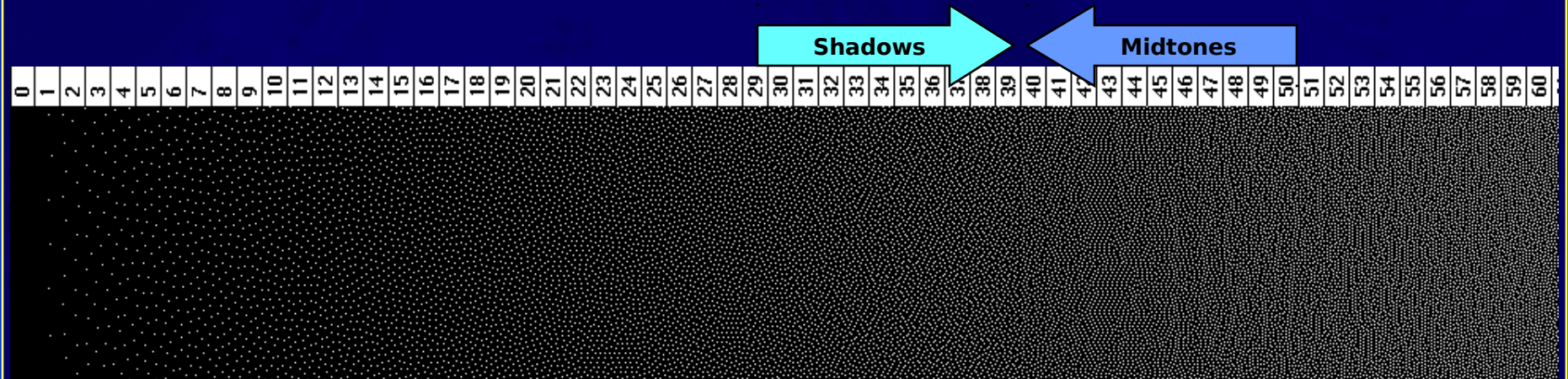
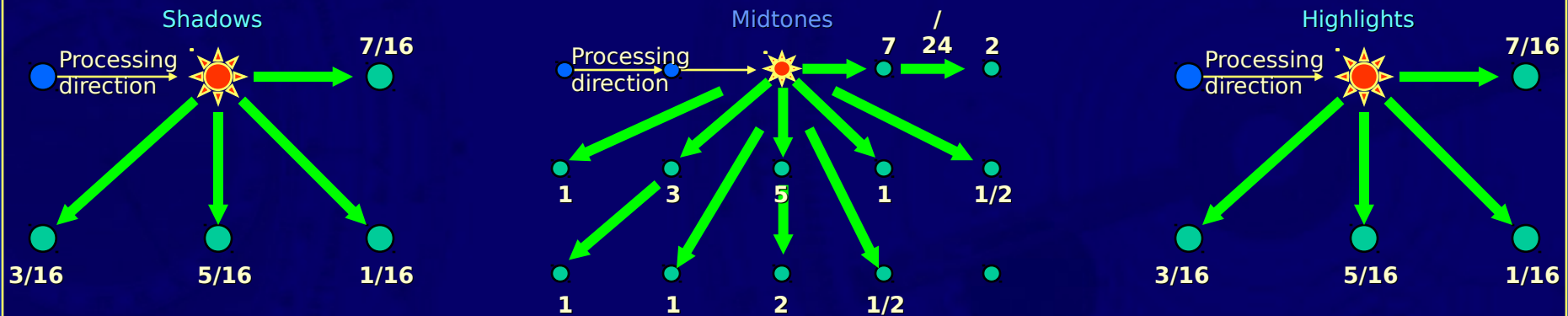


# Eschbach: Input-Dependent Weights, Independent Matrices (1993)





# Eschbach: Input-Dependent Weights, Dependent Matrices



# Eschbach: Input-Dependent Weights

## Conclusions

- **Visual Artifacts can be improved Using Input-Dependent Weights**
- **Correlation Between Matrices May Reduce Banding Effect**

## Weak Points of the Method

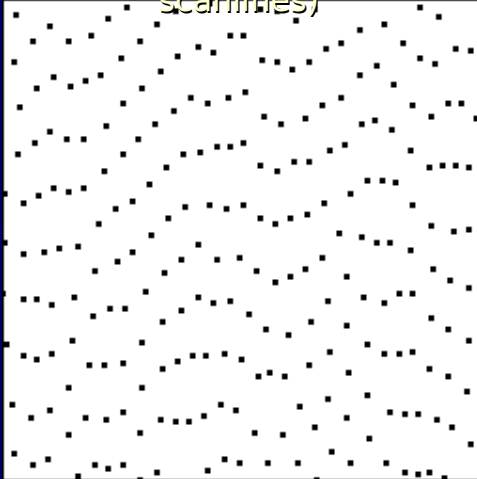
- **No Objective Criterion on Input-Dependent Weights**
- **No Mechanism Proposed To Insure Correlation Between Weights**

## Our Contribution

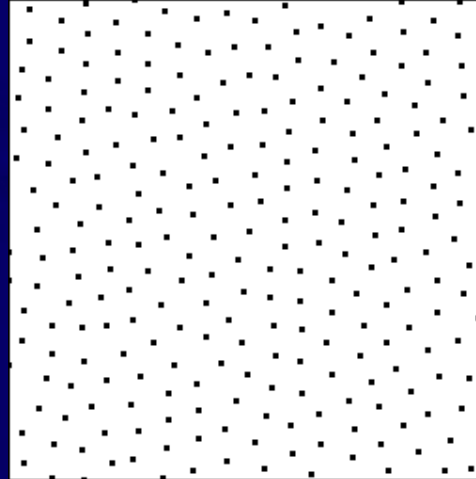
- 1. Provide Objective Criterion on Input-Dependent Weights**
- 2. Provide Mechanism To Insure Correlation Between Weights**
- 3. Implement an Error-Diffusion Algorithm**
  - **Conceptually Simple**
  - **Computationally Efficient**
  - **Almost Artifact-Free**
  - **Publicly Available**

# (1) Objective Criterion of Visual Quality of Halftoning

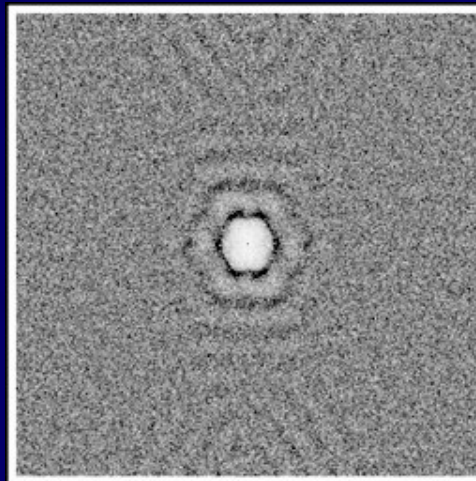
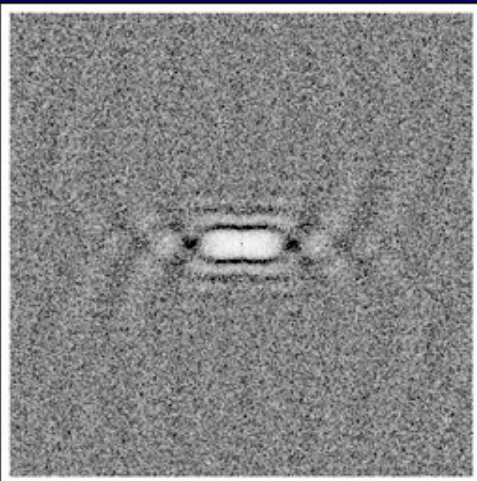
Bad Error-Diffusion  
(Std. Floyd-Steinberg,  
scanlines)



Good Error-Diffusion



Images

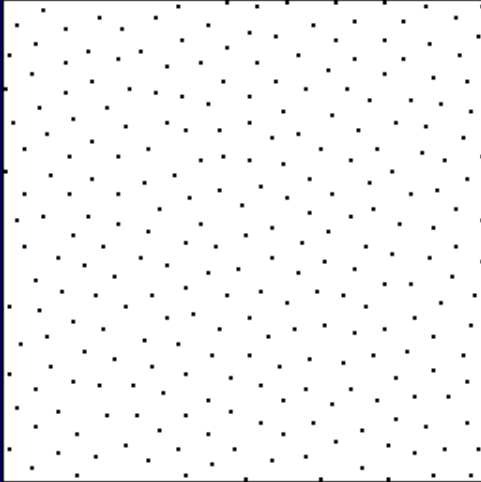


Fourier  
Amplitude Spectra

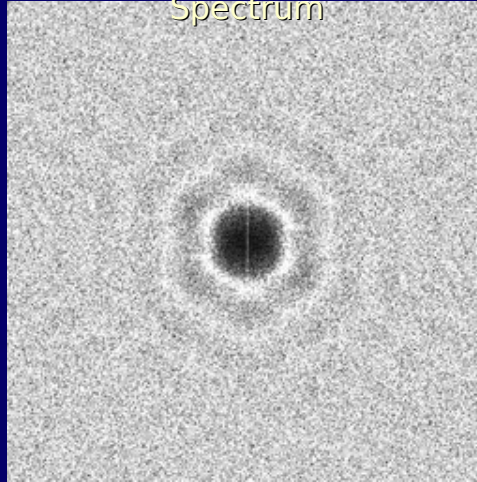


# (1) Objective Criterion of Visual Quality of Halftoning

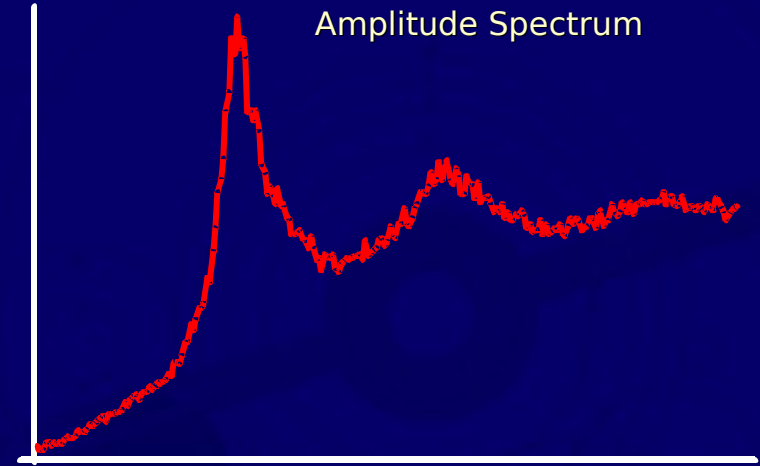
Image



Fourier Amplitude Spectrum

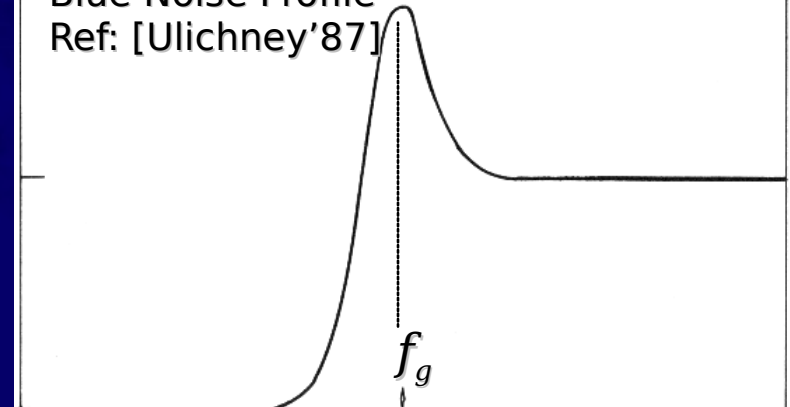


Profile of Fourier Amplitude Spectrum

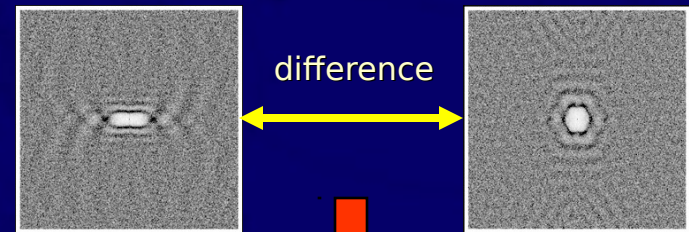
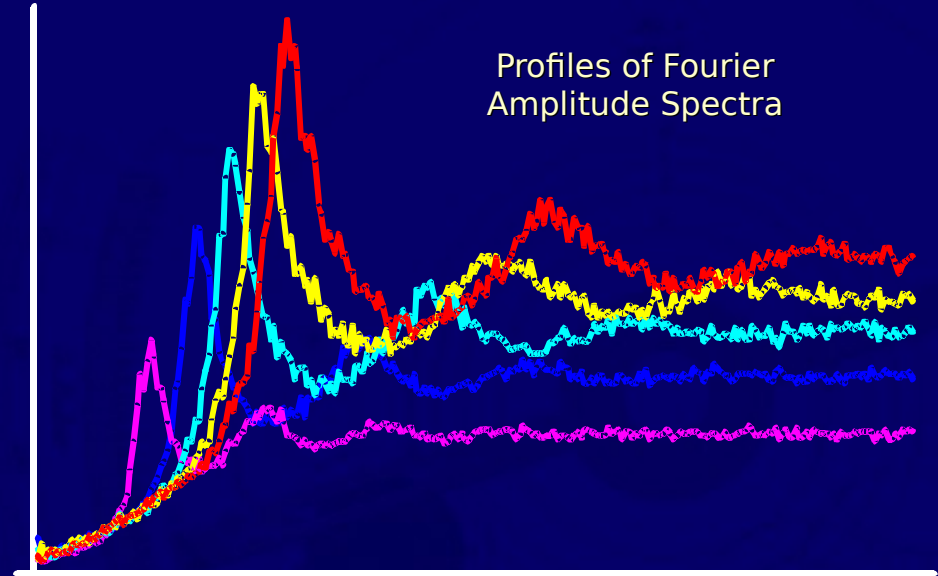
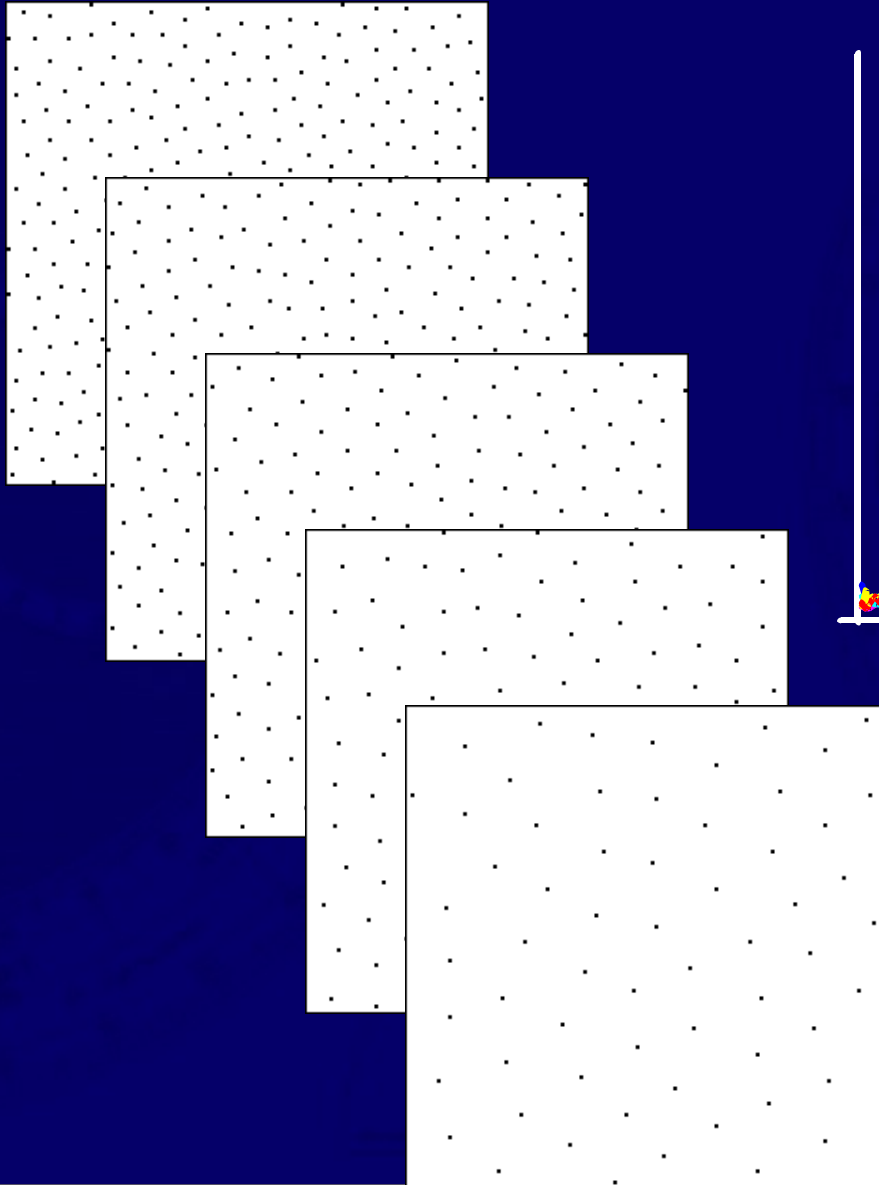


- **Radial Symmetry**
- **Characteristic Blue Noise Profile**

Blue Noise Profile  
Ref: [Ulichney'87]



# (1) Objective Criterion of Visual Quality of Halftoning

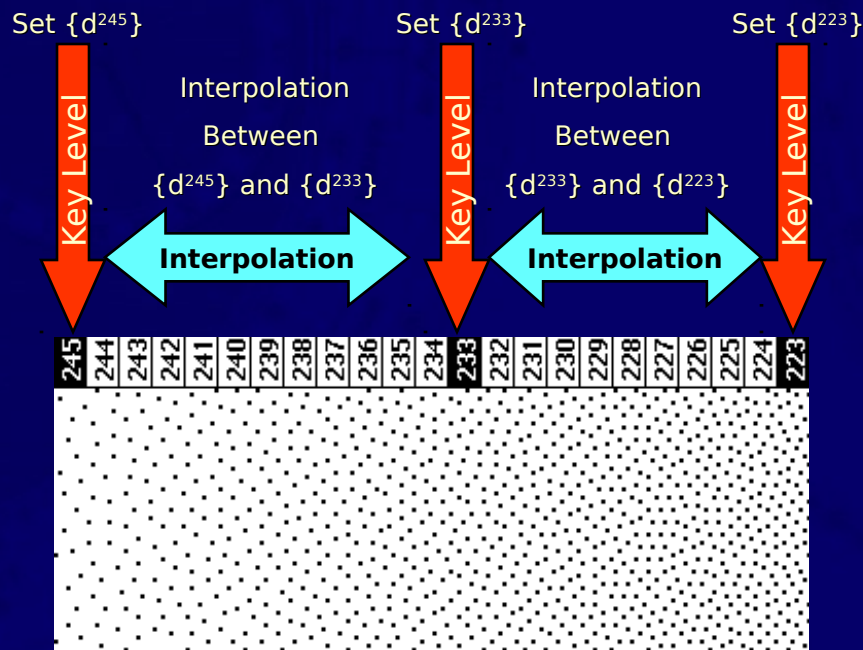


number  
optimization

## (2) Mechanism of Correlation Between E-D Weights

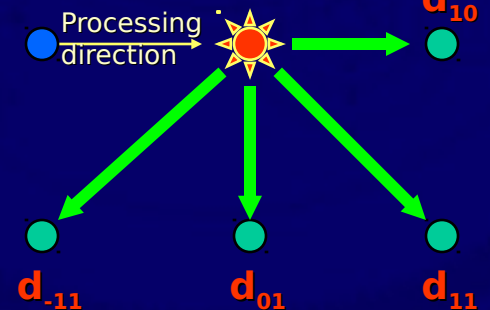


Key Intensity Levels



For each intensity level  $i$ ,  
The set  $\{d^i\}$  is defined  
as

$$\{d^i\} = \{d_{10}, d_{-11}, d_{01}, d_{11}\}$$





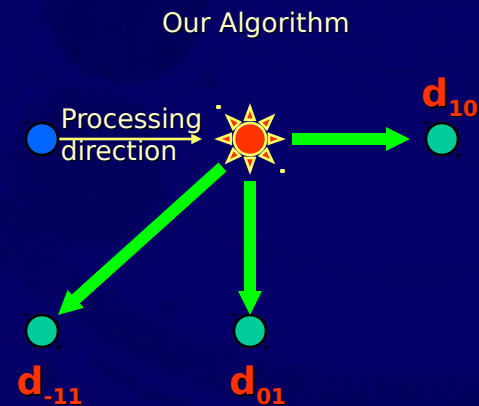
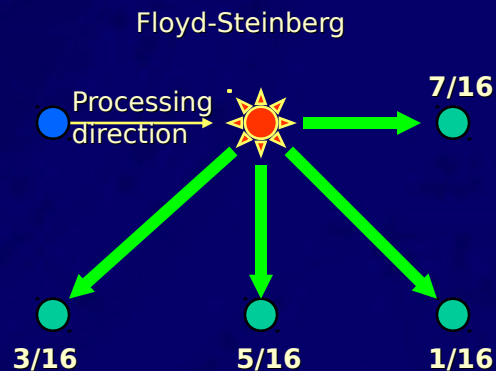
# Our Error-Diffusion Algorithm

## Assumptions:

- A. E-D Algorithm is good when its Fourier Amplitude Spectrum is close to “Blue Noise”
- B. Smooth behavior of E-D algorithm when coefficients  $\{d_i\}$  vary smoothly across the dynamic range of intensity levels  $\{i\}$
- C. Intensity levels 0, **1/4, 1/3, 1/2, 2/3, 3/4, 1** are potentially problematic: they may be source of artifacts
- D. Sets  $\{d_i\}$  and  $\{d_{255-i}\}$  generate almost identical artifacts structures, with white and black inverted

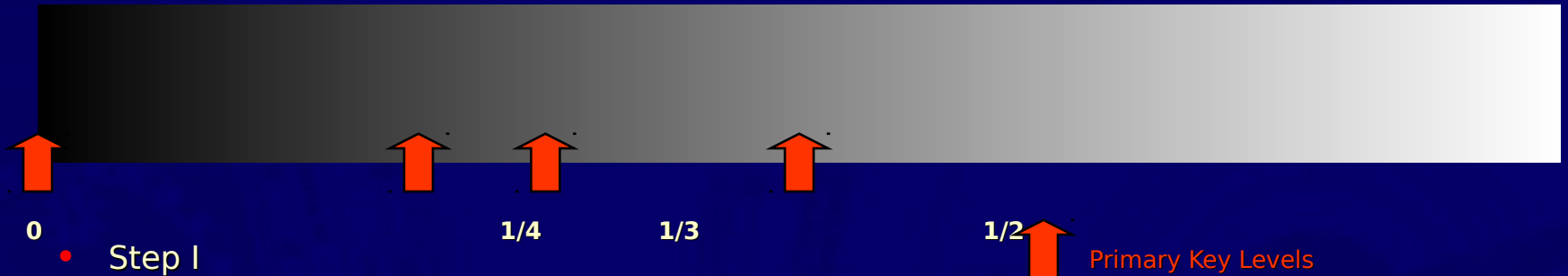
# Our Error-Diffusion Algorithm

- **Simplified Set of Distribution Coefficients**  $d^i = \{d_{10}, d_{-11}, d_{01}\}$  for Each Intensity Level  $i$ 
  - **Faster**
  - **Easier to Optimize**
  - **Sufficient for Achieving Good Quality**



# Our Error-Diffusion Algorithm: Finding $\{d_{10}, d_{-11}, d_{01}\}$

Given: Initial set of key level between 0 and  $\frac{1}{2}$ , where artifacts are *a priori* present

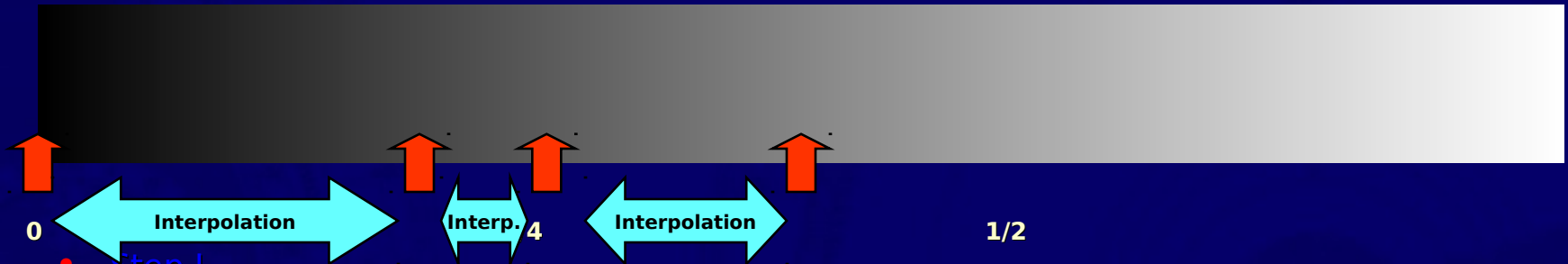


For each key level, find  $d_{key} = \{d_{10}, d_{-11}, d_{01}\}$  that approaches “Blue Noise”, by minimizing the difference between given Fourier Amplitude Spectrum and Ideal “Blue Noise Profile”



# Our Error-Diffusion Algorithm: Finding $\{d_{10}, d_{-11}, d_{01}\}$

Given: Initial set of key level between 0 and  $\frac{1}{2}$ , where artifacts are *a priori* present



- Step I

For each key level, find  $d^{key} = \{d_{10}, d_{-11}, d_{01}\}$  that approaches “Blue Noise”, by minimizing the difference between given Fourier Amplitude Spectrum and Ideal “Blue Noise Profile”

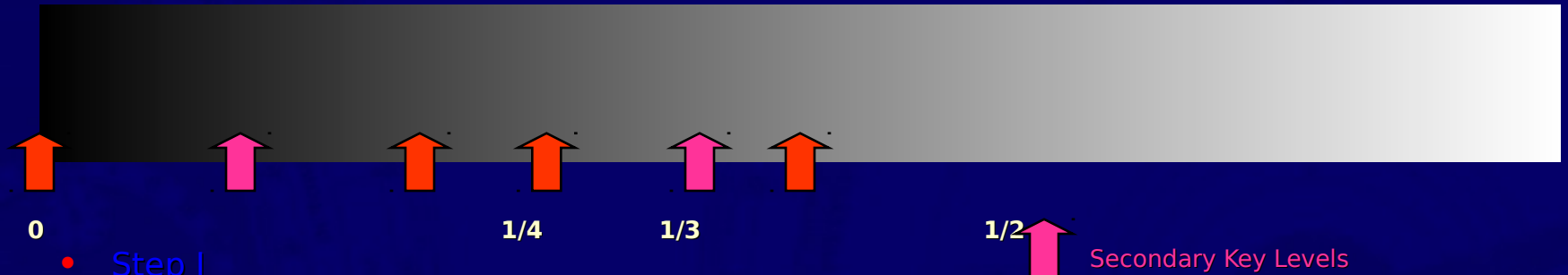
- Step II

Interpolate between key level.  
Visually check for artifacts in-between.

0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61 62 63

# Our Error-Diffusion Algorithm: Finding $\{d_{10}, d_{-11}, d_{01}\}$

Given: Initial set of key level between 0 and  $\frac{1}{2}$ , where artifacts are *a priori* present



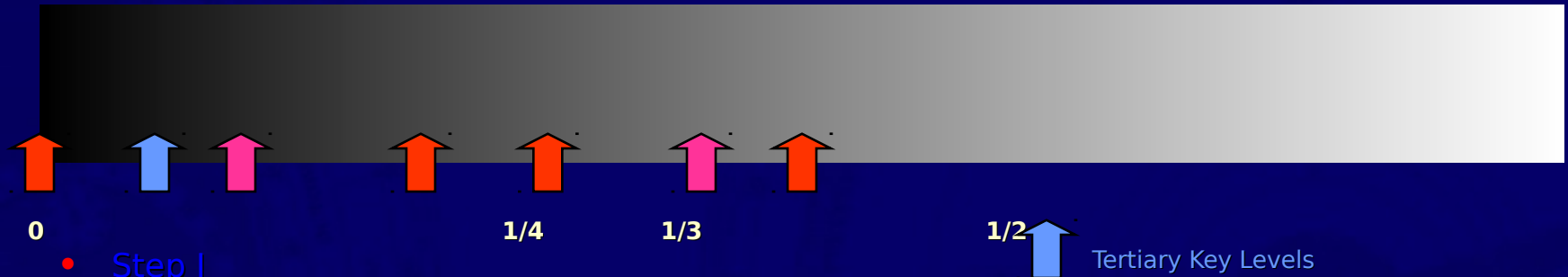
- Step I  
For each key level, find  $d^{key}=\{d_{10}, d_{-11}, d_{01}\}$  that approaches “Blue Noise”, by minimizing the difference between given Fourier Amplitude Spectrum and Ideal “Blue Noise Profile”
- Step II  
Interpolate between key level.  
Visually check for artifacts in-between.  
If not OK,  
define new key levels and jump to Step I  
otherwise:  
continue

0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61 62 63

a

# Our Error-Diffusion Algorithm: Finding $\{d_{10}, d_{-11}, d_{01}\}$

Given: Initial set of key level between 0 and  $\frac{1}{2}$ , where artifacts are *a priori* present

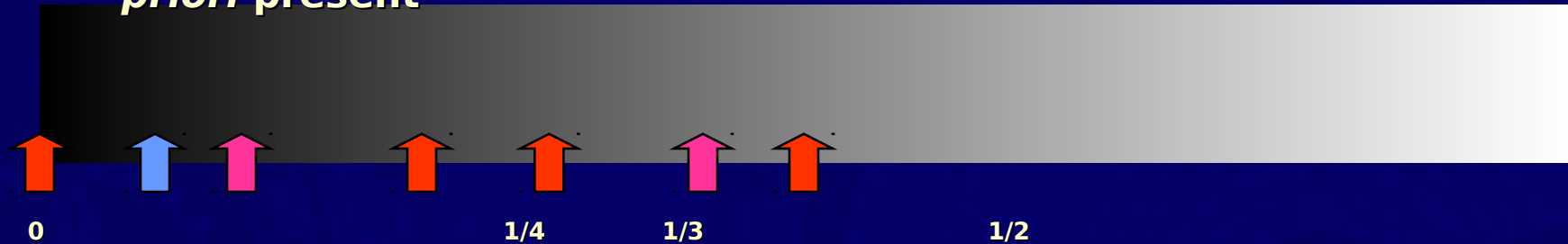


- Step I  
For each key level, find  $d^{key} = \{d_{10}, d_{-11}, d_{01}\}$  that approaches “Blue Noise”, by minimizing the difference between given Fourier Amplitude Spectrum and Ideal “Blue Noise Profile”
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# Our Error-Diffusion Algorithm: Finding $\{d_{10}, d_{-11}, d_{01}\}$

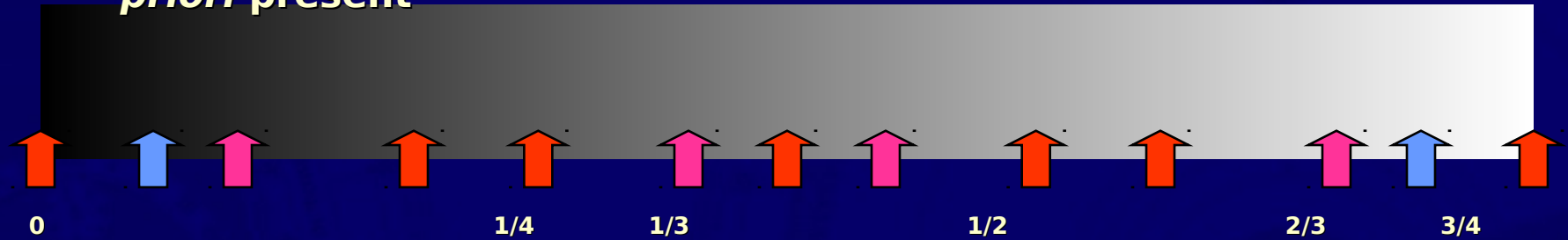
**Given: Initial set of key level between 0 and  $\frac{1}{2}$ , where artifacts are a *priori* present**



- **Step I**  
 For each key level, find  $d^{key} = \{d_{10}, d_{-11}, d_{01}\}$  that approaches “Blue Noise”, by minimizing the difference between given Fourier Amplitude Spectrum and Ideal “Blue Noise Profile”
- **Step II**  
 Interpolate between key level.  
 Visually check for artifacts in-between.  
 If not OK,  
     define new key levels and jump to Step I  
 otherwise:  
     continue
- **Step III**  
 Extend Solution Symmetrically, about  $\frac{1}{2}$

# Our Error-Diffusion Algorithm: Finding $\{d_{10}, d_{-11}, d_{01}\}$

**Given: Initial set of key level between 0 and  $\frac{1}{2}$ , where artifacts are a *priori* present**



- **Step I**  
For each key level, find  $d^{key} = \{d_{10}, d_{-11}, d_{01}\}$  that approaches “Blue Noise”, by minimizing the difference between given Fourier Amplitude Spectrum and Ideal “Blue Noise Profile”
- **Step II**  
Interpolate between key level.  
Visually check for artifacts in-between.  
If not OK,  
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continue
- **Step III**  
Extend Solution Symmetrically, about  $\frac{1}{2}$

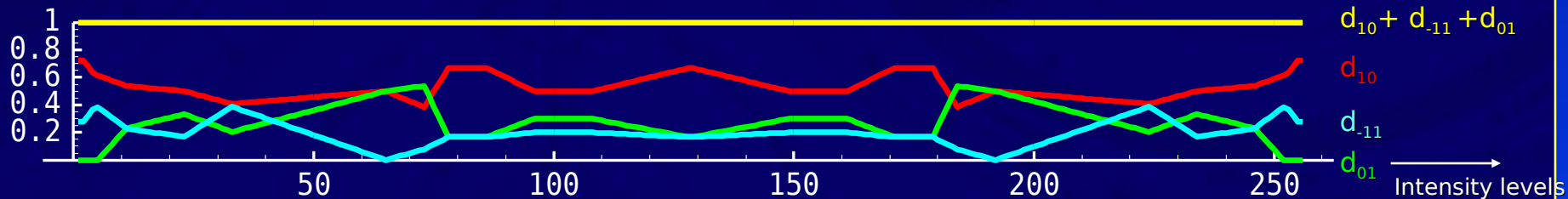
## Finding $\{d_{10}, d_{-11}, d_{01}\}$ : Results

**0:** 13, 0, 5  
**1:** 13, 0, 5  
**2:** 21, 0, 10  
**3:** 7, 0, 4  
**4:** 8, 0, 5  
**5:** 47, 3, 28  
**6:** 23, 3, 13  
**7:** 15, 3, 8  
**8:** 22, 6, 11  
**9:** 43, 15, 20  
**10:** 7, 3, 3  
**11:** 501, 224, 211  
**12:** 249, 116, 103  
**13:** 165, 80, 67  
**14:** 123, 62, 49  
**15:** 489, 256, 191  
**16:** 81, 44, 31  
**17:** 483, 272, 181  
**18:** 60, 35, 22  
**19:** 53, 32, 19  
**20:** 237, 148, 83  
**21:** 471, 304, 161  
**22:** 3, 2, 1  
**23:** 481, 314, 185  
**24:** 354, 226, 155  
**25:** 1389, 866, 685  
**26:** 227, 138, 125  
**27:** 267, 158, 163  
**28:** 327, 188, 220  
**29:** 61, 34, 45  
**30:** 627, 338, 505  
**31:** 1227, 638, 1075

**32:** 20, 10, 19  
**33:** 1937, 1000, 1767  
**34:** 977, 520, 855  
**35:** 657, 360, 551  
**36:** 71, 40, 57  
**37:** 2005, 1160, 1539  
**38:** 337, 200, 247  
**39:** 2039, 1240, 1425  
**40:** 257, 160, 171  
**41:** 691, 440, 437  
**42:** 1045, 680, 627  
**43:** 301, 200, 171  
**44:** 177, 120, 95  
**45:** 2141, 1480, 1083  
**46:** 1079, 760, 513  
**47:** 725, 520, 323  
**48:** 137, 100, 57  
**49:** 2209, 1640, 855  
**50:** 53, 40, 19  
**51:** 2243, 1720, 741  
**52:** 565, 440, 171  
**53:** 759, 600, 209  
**54:** 1147, 920, 285  
**55:** 2311, 1880, 513  
**56:** 97, 80, 19  
**57:** 335, 280, 57  
**58:** 1181, 1000, 171  
**59:** 793, 680, 95  
**60:** 599, 520, 57  
**61:** 2413, 2120, 171  
**62:** 405, 360, 19  
**63:** 2447, 2200, 57

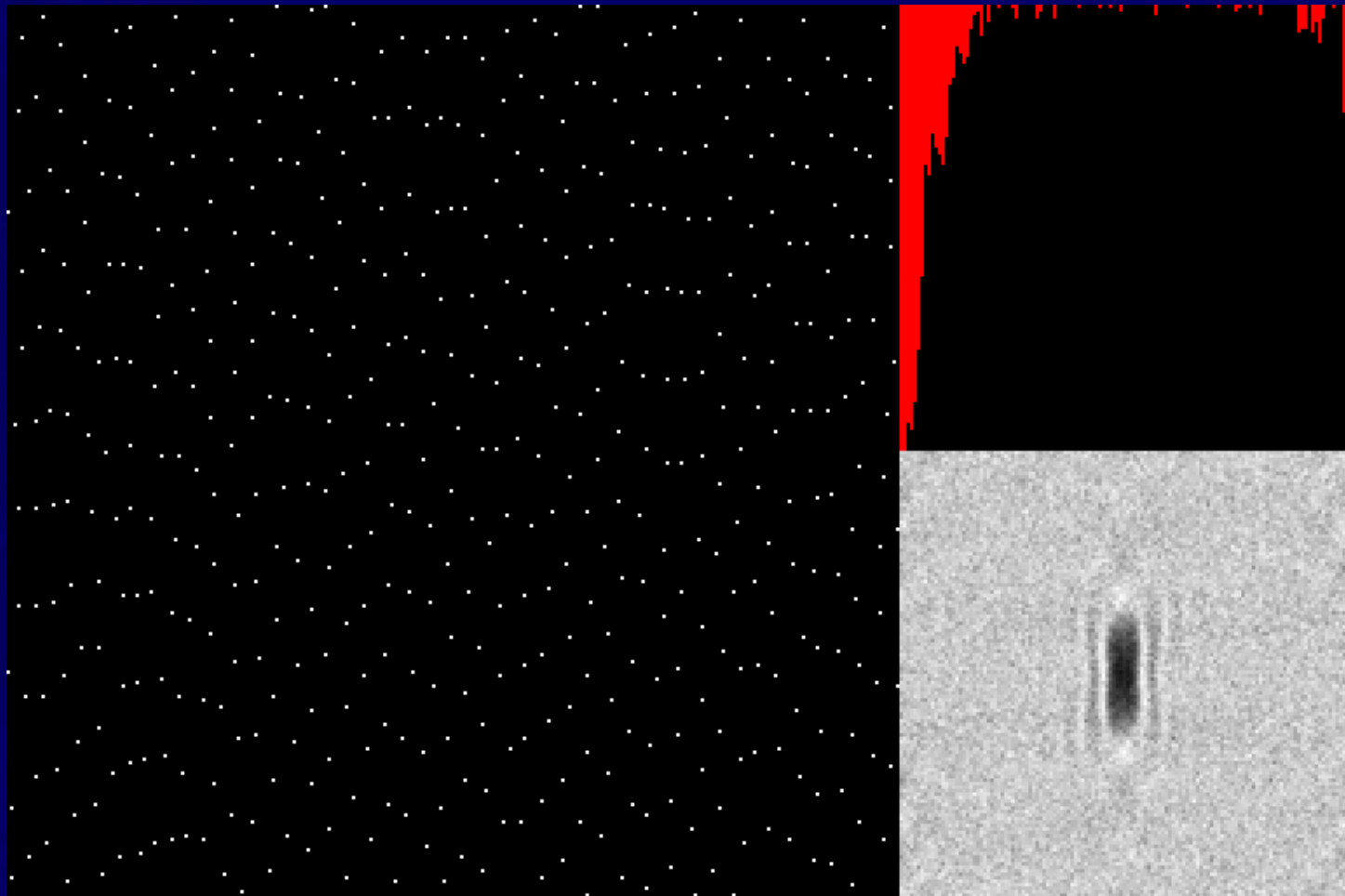
**64:** 11, 10, 0  
**65:** 158, 151, 3  
**66:** 178, 179, 7  
**67:** 1030, 1091, 63  
**68:** 248, 277, 21  
**69:** 318, 375, 35  
**70:** 458, 571, 63  
**71:** 878, 1159, 147  
**72:** 5, 7, 1  
**73:** 172, 181, 37  
**74:** 97, 76, 22  
**75:** 72, 41, 17  
**76:** 119, 47, 29  
**77:** 4, 1, 1  
**78:** 4, 1, 1  
**79:** 4, 1, 1  
**80:** 4, 1, 1  
**81:** 4, 1, 1  
**82:** 4, 1, 1  
**83:** 4, 1, 1  
**84:** 4, 1, 1  
**85:** 4, 1, 1  
**86:** 65, 18, 17  
**87:** 95, 29, 26  
**88:** 185, 62, 53  
**89:** 30, 11, 9  
**90:** 35, 14, 11  
**91:** 85, 37, 28  
**92:** 55, 26, 19  
**93:** 80, 41, 29  
**94:** 155, 86, 59  
**95:** 5, 3, 2

**96:** 5, 3, 2  
**97:** 5, 3, 2  
**98:** 5, 3, 2  
**99:** 5, 3, 2  
**100:** 5, 3, 2  
**101:** 5, 3, 2  
**102:** 5, 3, 2  
**103:** 5, 3, 2  
**104:** 5, 3, 2  
**105:** 5, 3, 2  
**106:** 5, 3, 2  
**107:** 5, 3, 2  
**108:** 305, 176, 119  
**109:** 155, 86, 59  
**110:** 105, 56, 39  
**111:** 80, 41, 29  
**112:** 65, 32, 23  
**113:** 55, 26, 19  
**114:** 335, 152, 113  
**115:** 85, 37, 28  
**116:** 115, 48, 37  
**117:** 35, 14, 11  
**118:** 355, 136, 109  
**119:** 30, 11, 9  
**120:** 365, 128, 107  
**121:** 185, 62, 53  
**122:** 25, 8, 7  
**123:** 95, 29, 26  
**124:** 385, 112, 103  
**125:** 65, 18, 17  
**126:** 395, 104, 101  
**127:** 4, 1, 1



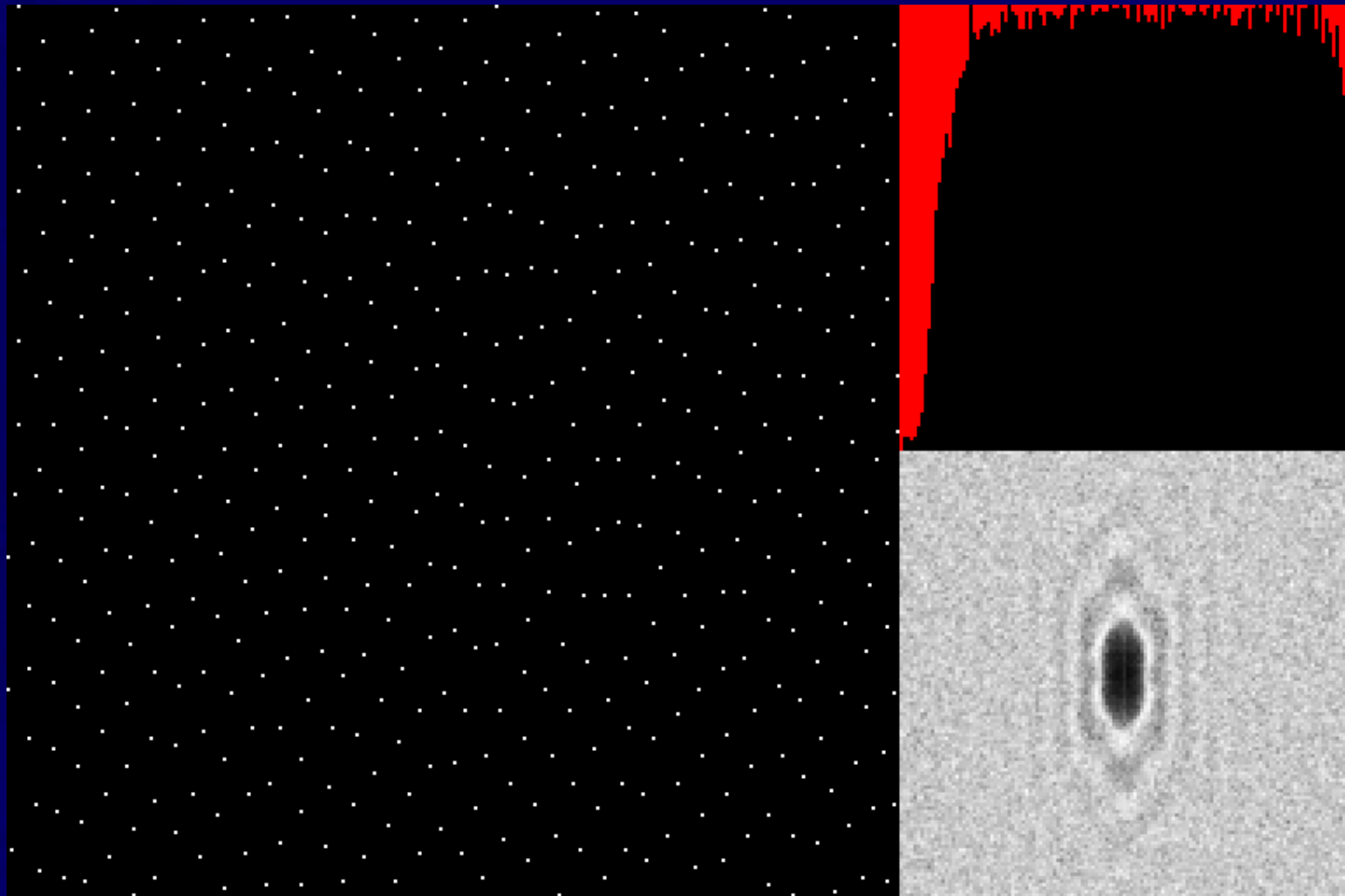


## Step I, Iteration # 1



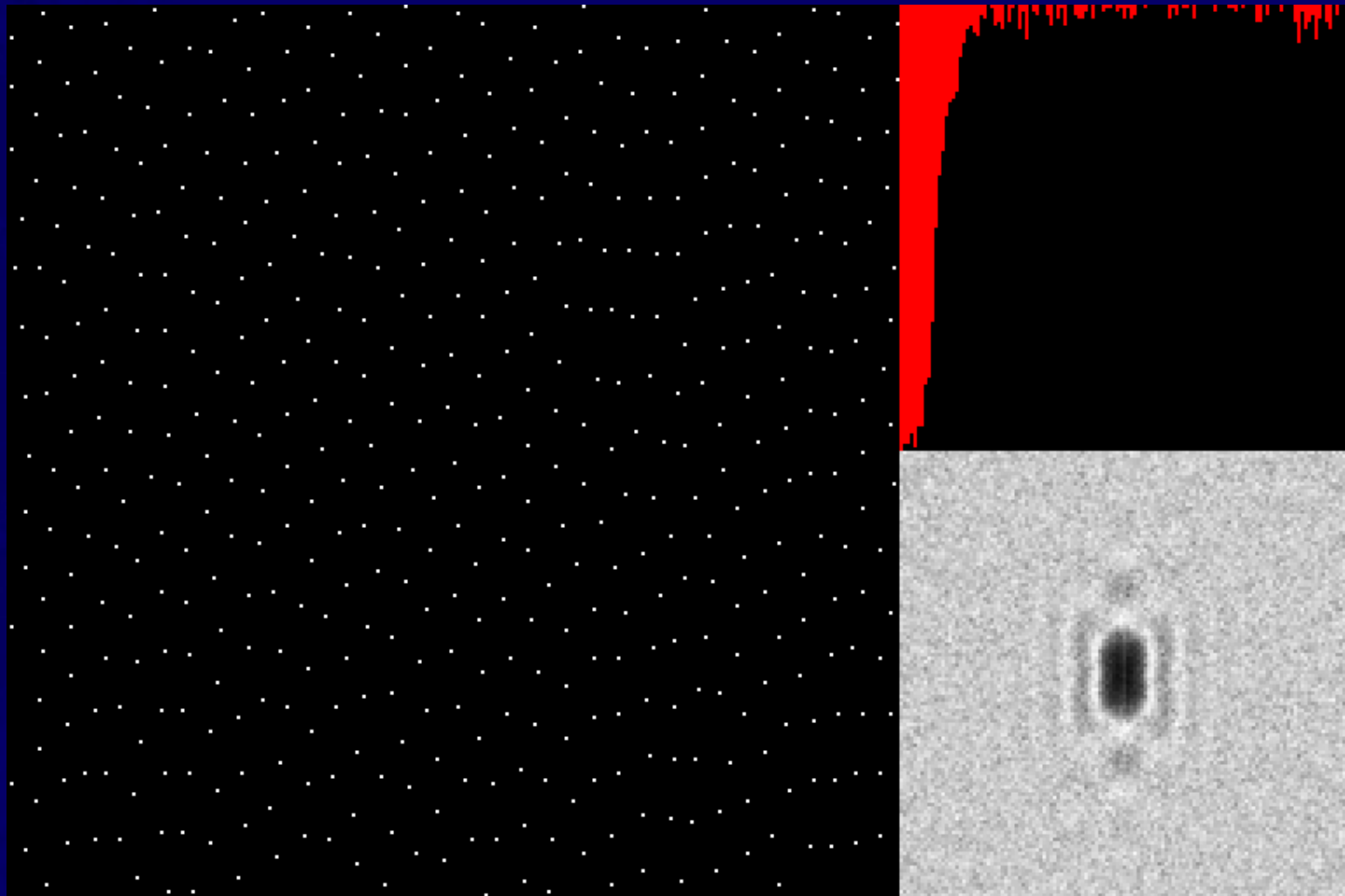
$$\begin{aligned} d_{10} &= 0.333333 \\ d_{-11} &= 0.333333 \\ d_{01} &= 0.333333 \end{aligned}$$

## Step I, Iteration # 3



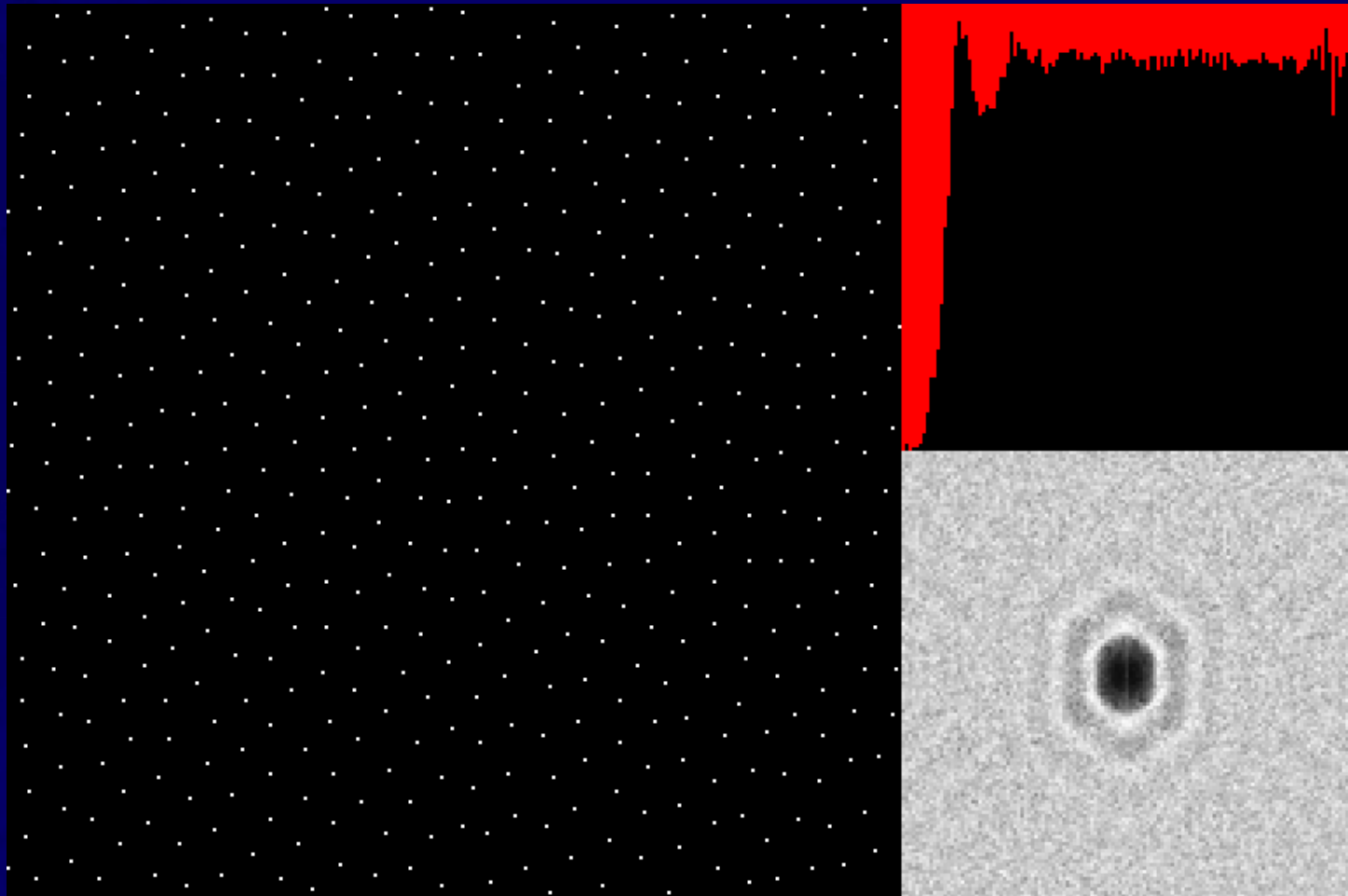
$d_{10} = 0.445736$   
 $d_{-11} = 0.306202$   
 $d_{01} = 0.248062$

## Step I, Iteration # 7



$$\begin{aligned} d_{10} &= 0.527778 \\ d_{-11} &= 0.444445 \\ d_{01} &= 0.027776 \end{aligned}$$

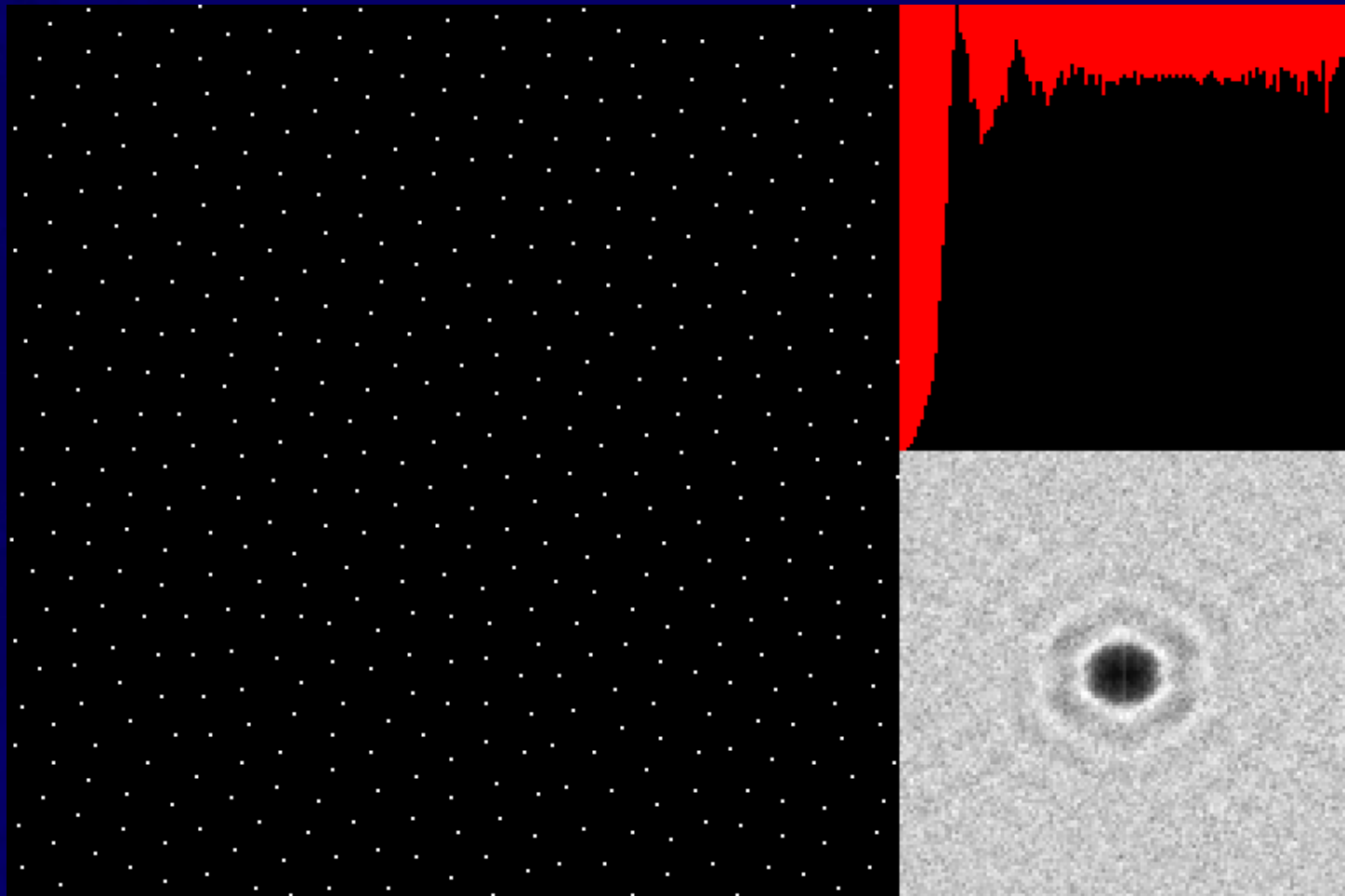
## Step I, Iteration # 12



$d_{10} = 0.637344$   
 $d_{-11} = 0.186967$   
 $d_{01} = 0.175689$

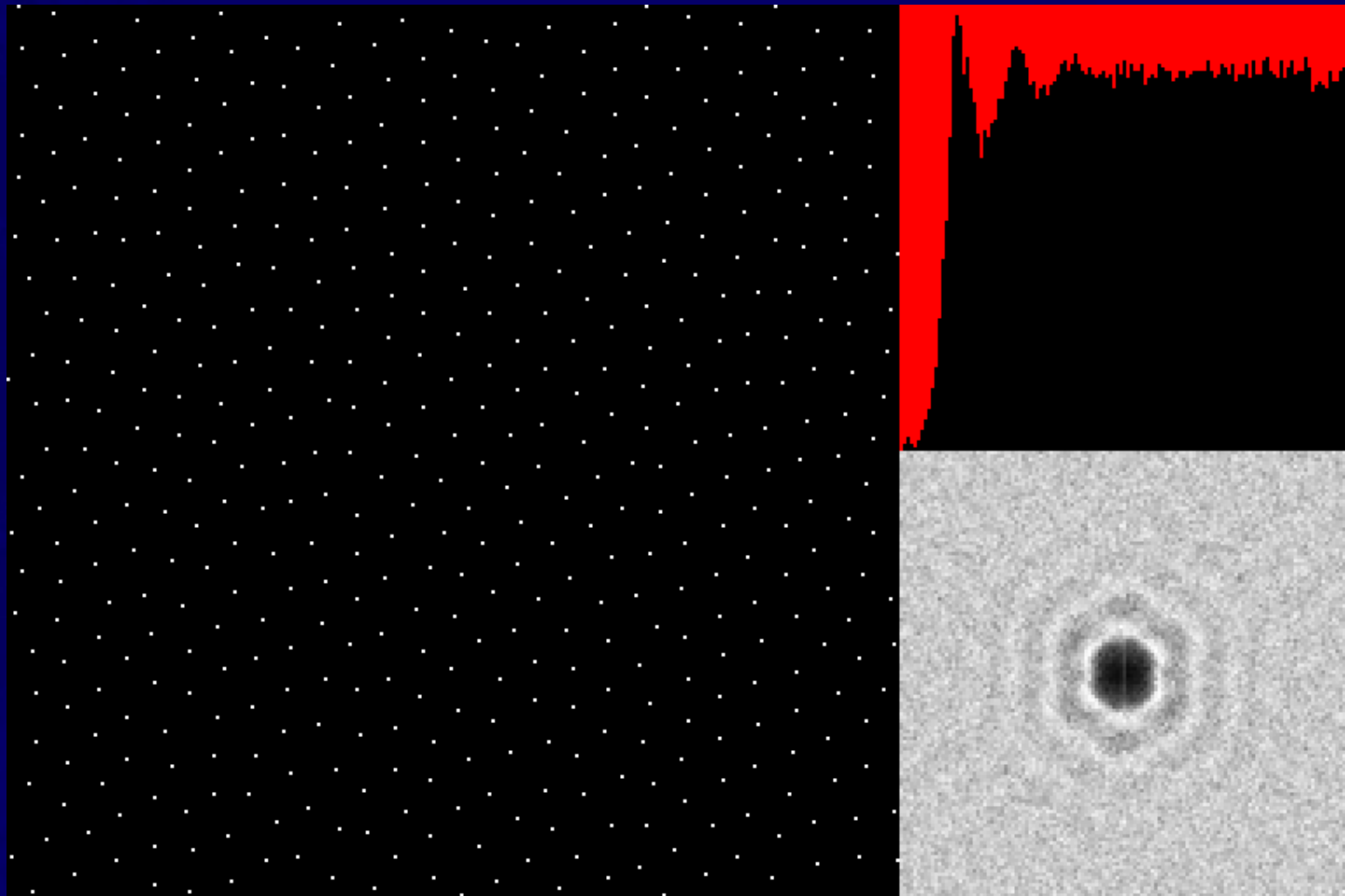


## Step I, Iteration # 15



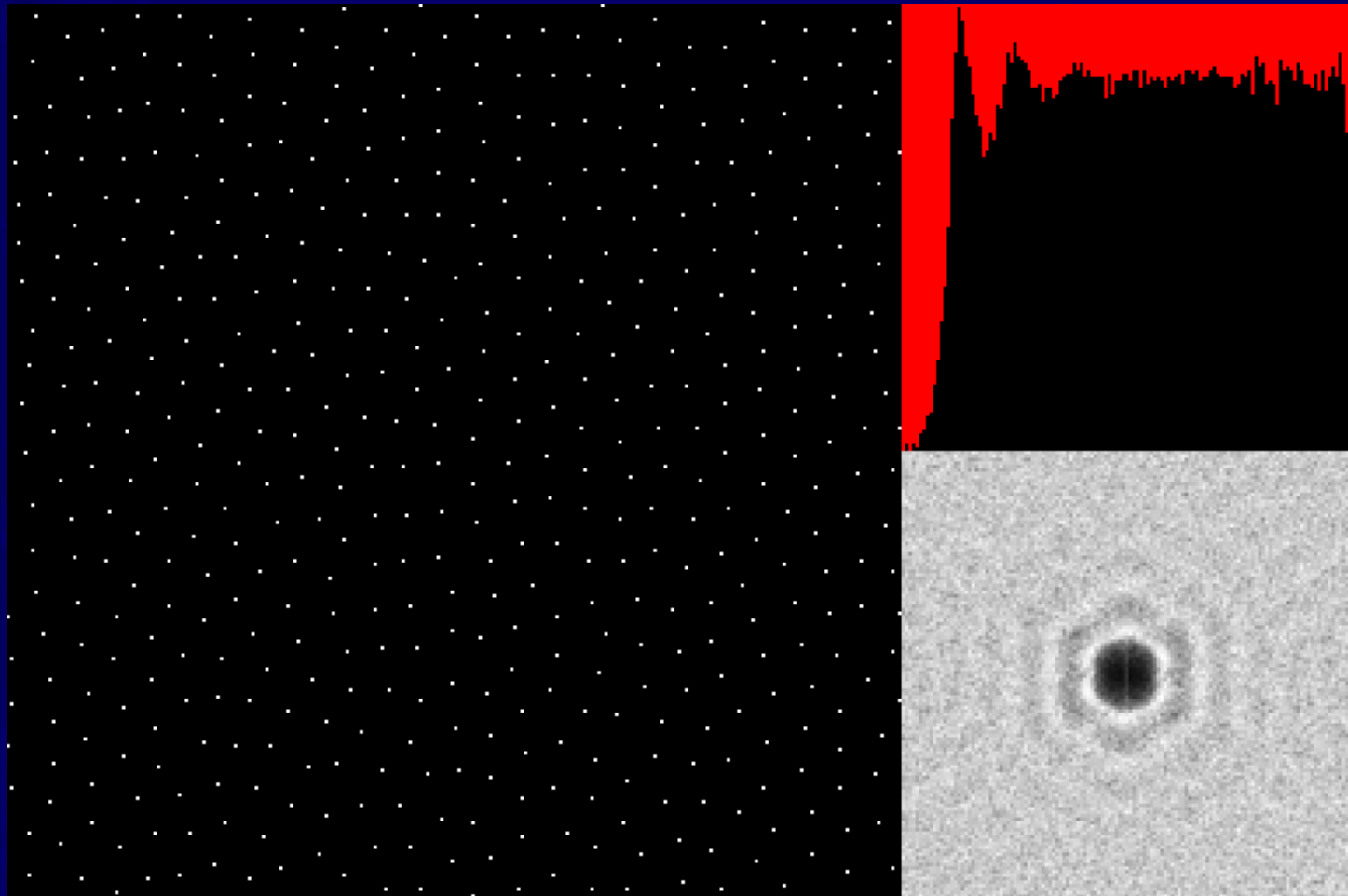
$d_{10} = 0.630038$   
 $d_{-11} = 0.218744$   
 $d_{01} = 0.151219$

## Step I, Iteration # 20



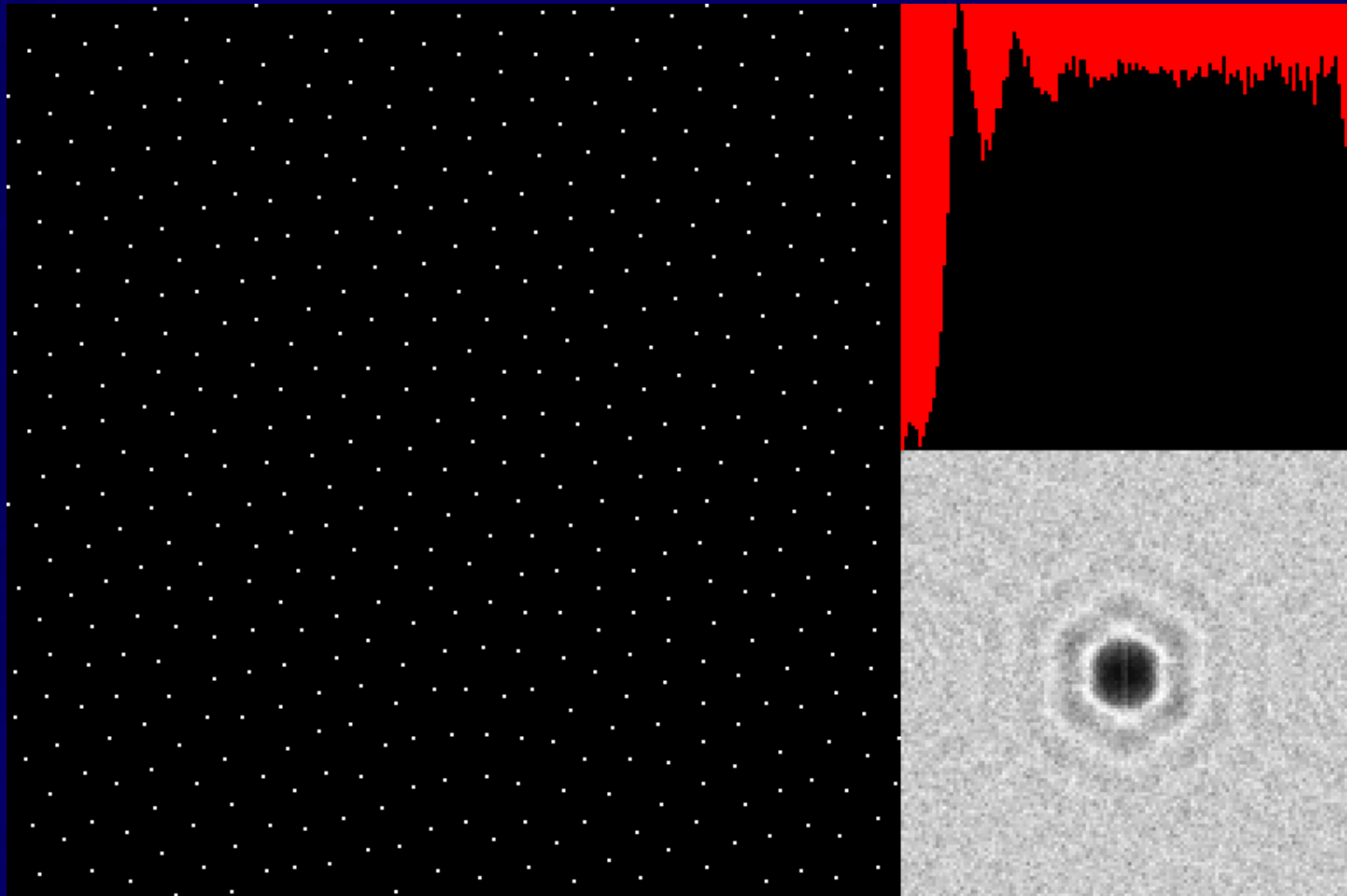
$$\begin{aligned} d_{10} &= 0.673123 \\ d_{11} &= 0.114484 \\ d_{01} &= 0.212393 \end{aligned}$$

## Step I, Iteration # 33



$d_{10} = 0.664381$   
 $d_{-11} = 0.185621$   
 $d_{01} = 0.149998$

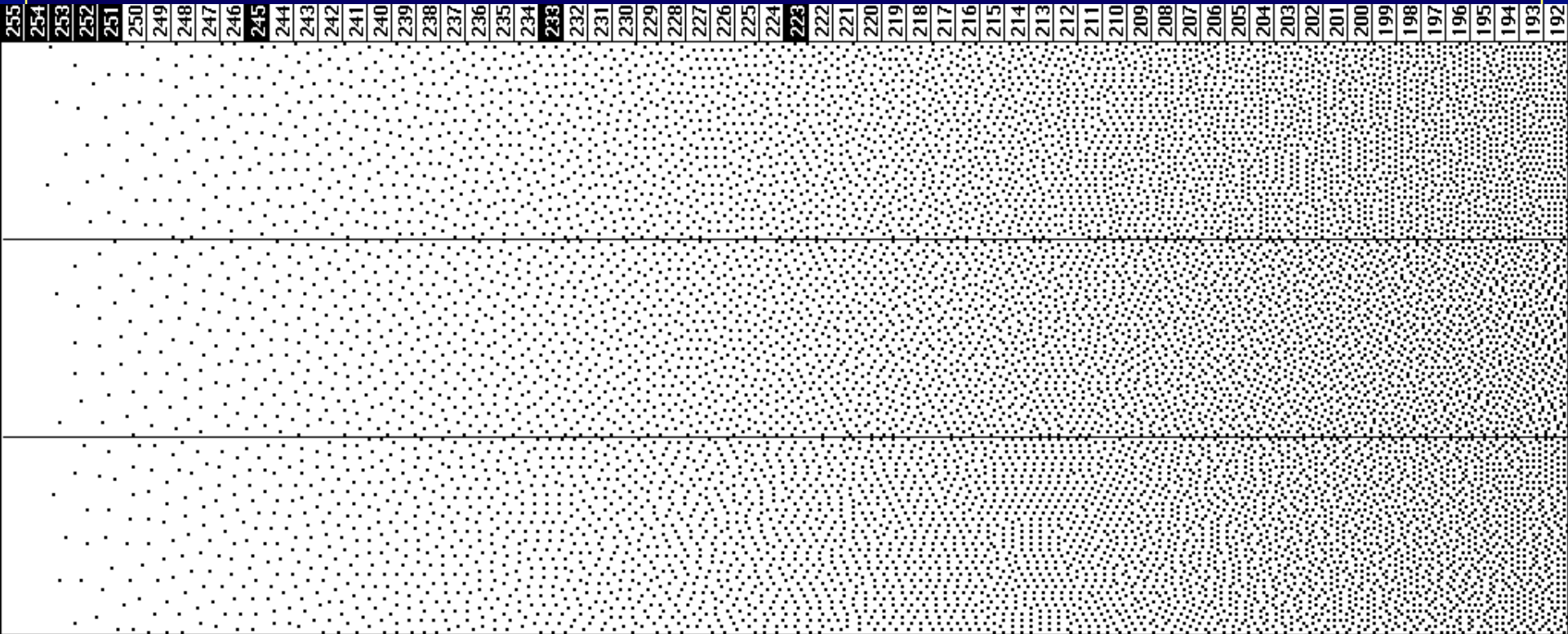
## Step I, Iteration # 41



$$\begin{aligned} d_{10} &= 0.655903 \\ d_{11} &= 0.182528 \\ d_{01} &= 0.16157 \end{aligned}$$

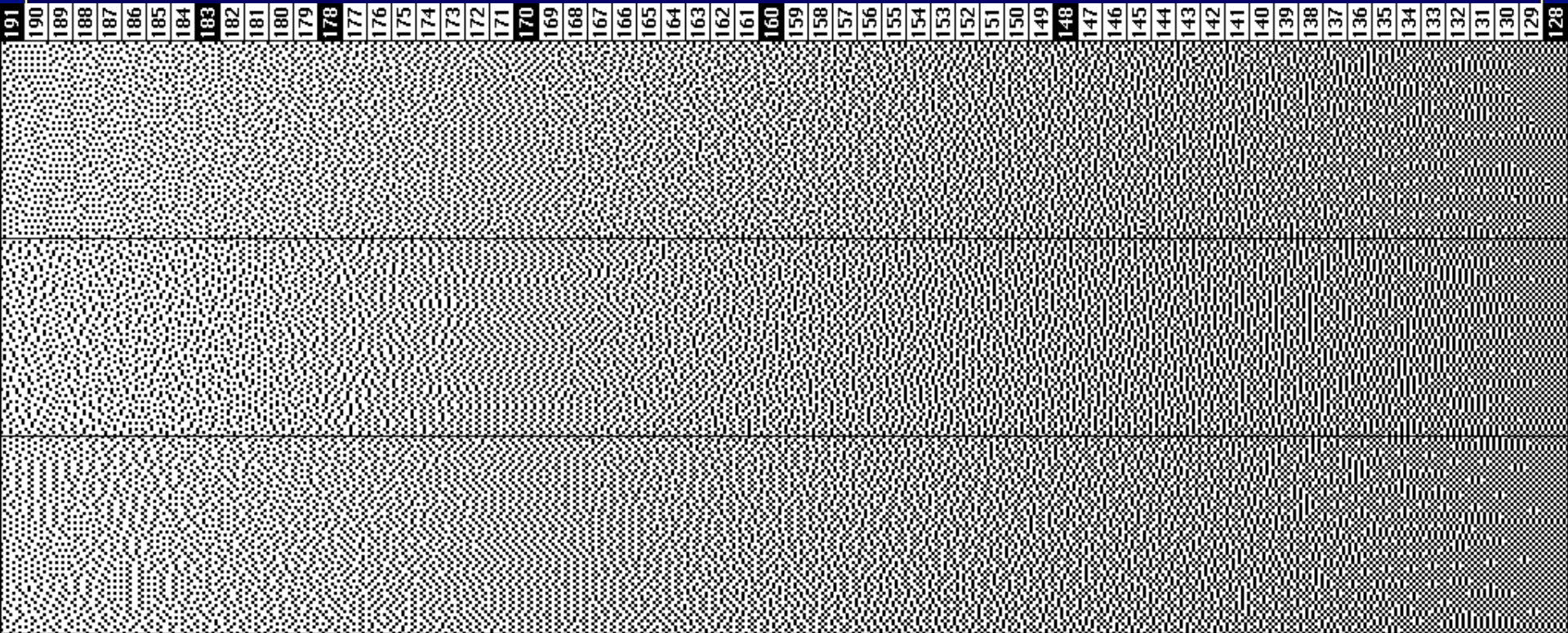


# Results



Top: Std. Floyd-Steinberg  
E-D  
Middle: Our Method  
Bottom: Shiau-Fan E-D

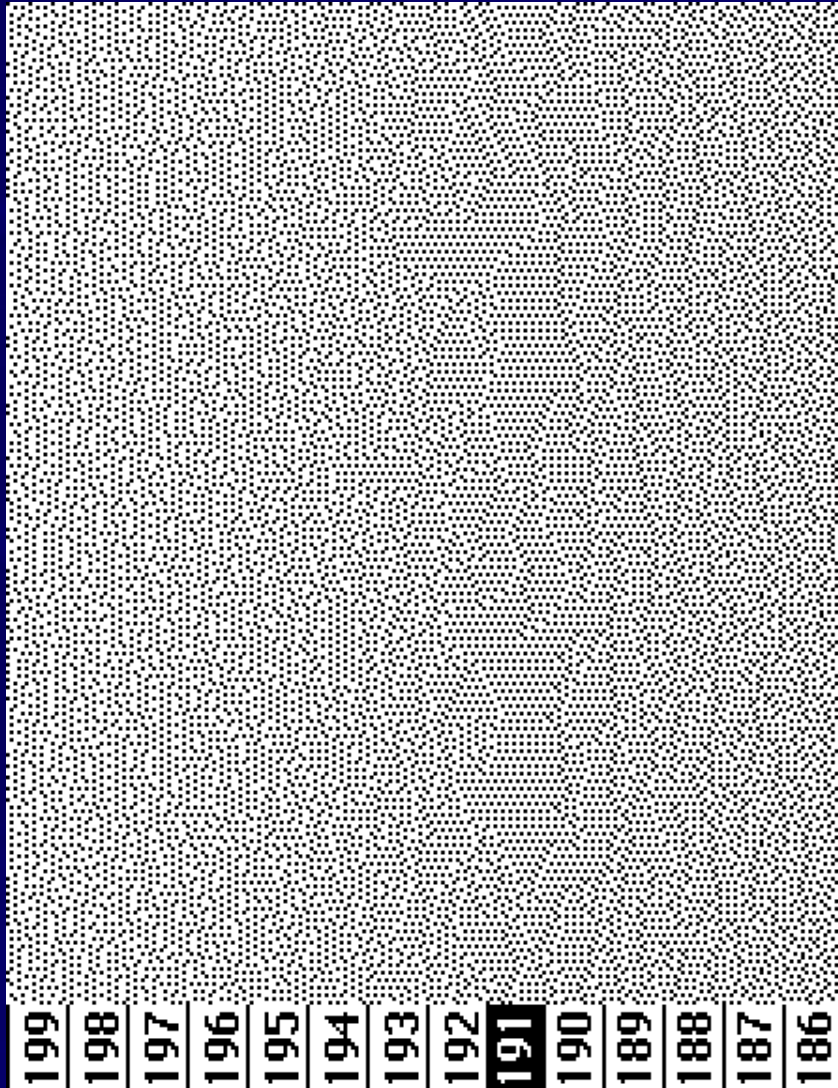
# Results



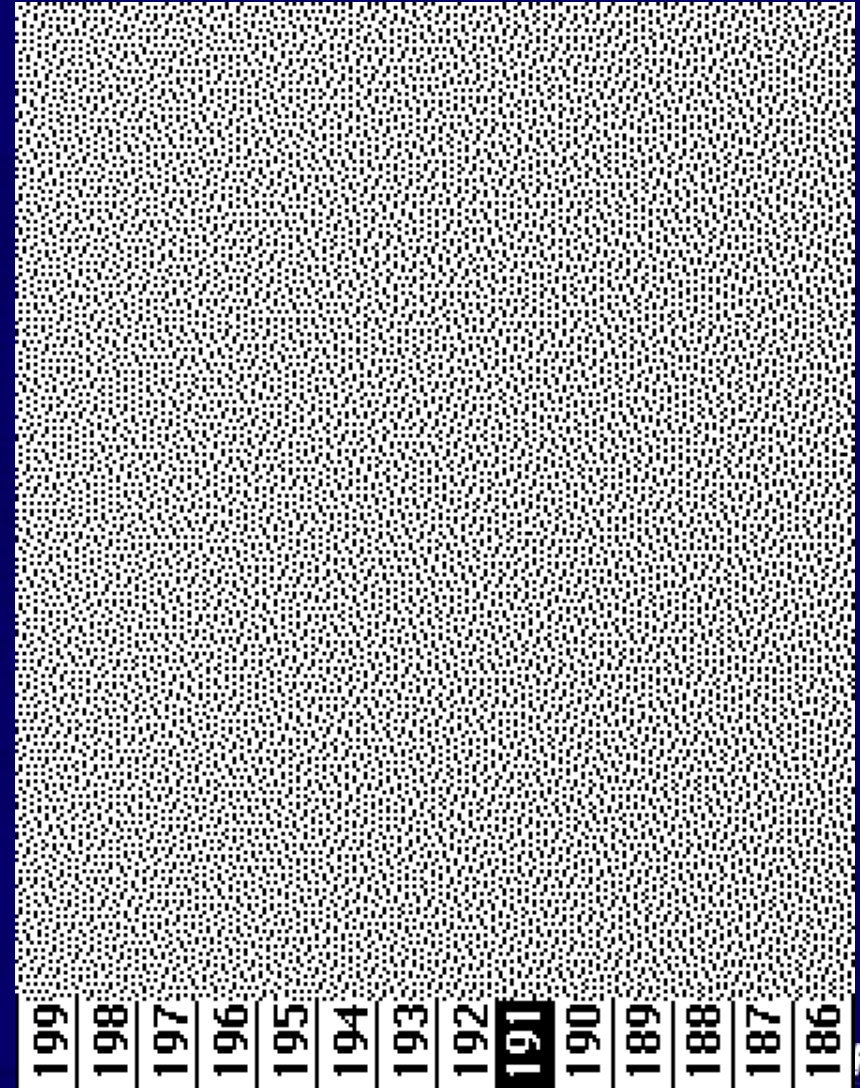
Top: Std. Floyd-Steinberg  
E-D  
Middle: Our Method  
Bottom: Shiau-Fan E-D

# Results

Std. Floyd-Steinberg E-D



Our Method





## Results

Monte-Carlo



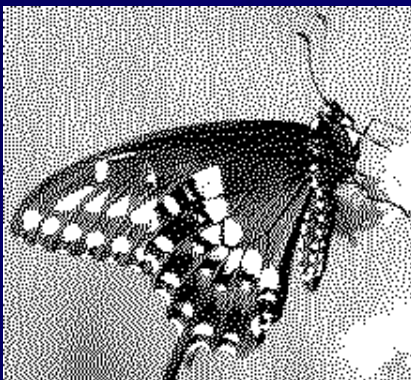
Our Method



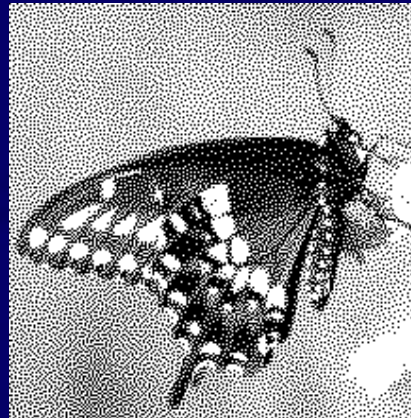
Clustered-Dot Dither



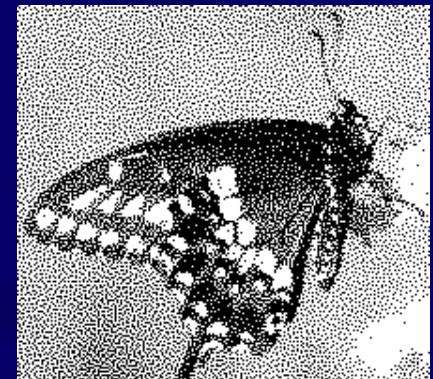
Floyd-Steinberg E-D



DBS-based



Blue Noise Mask





## Conclusions

- **High-speed Error-Diffusion algorithm**
- **Good visual quality**
- **Conceptually simple**
- **Publicly available on**

**<http://www.iro.umontreal.ca/~ostrom/varcoeffED>**

- **Results of further development will be placed at the same address:**

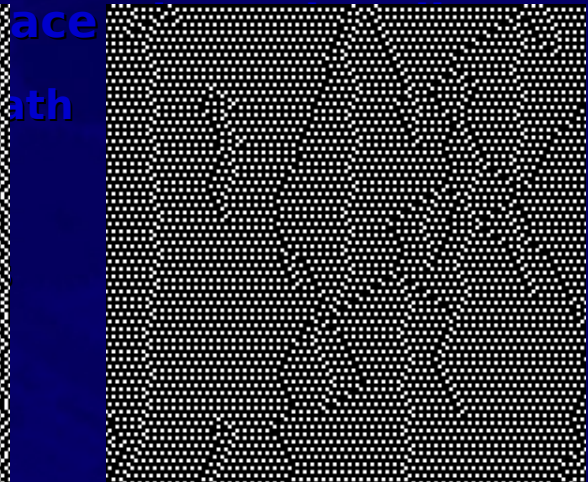
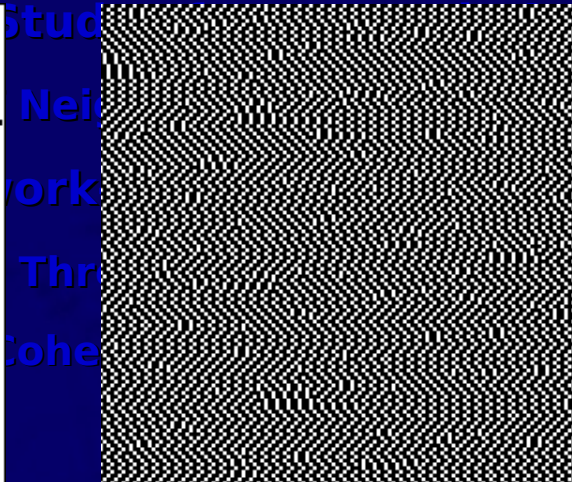
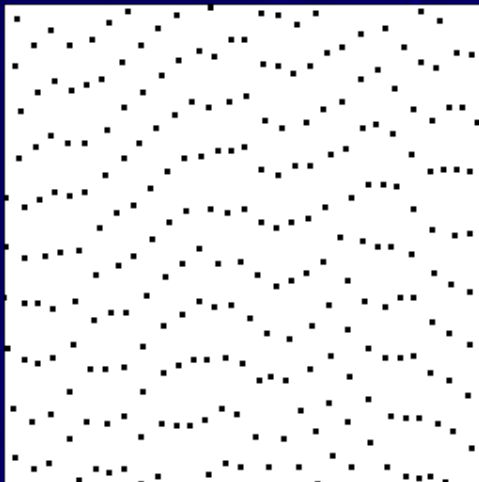
**<http://www.iro.umontreal.ca/~ostrom/varcoeffED>**

## Future Work

- **Taxonomy of Artifacts in E-D**
  - Different Artifacts need Different Processing
  - Current Work by P-M Jodoin: Case around  $\frac{1}{2}$
- **Better Cost Function (Automatic Quality Evaluation)**
  - Pathetic Cases (Blue Noise Criterion Does Not Work)
  - Detection of Local Structures (Wavelets, Gabor Functions etc.)
- **Systematic Study of Parametric Space Dimensionality**
  - Number of Neighbors, Processing Path
- **E-D for Network-Oriented Imaging**
  - Coherence Through Multiple Depths
  - Temporal Coherence with E-D

## Future Work

- **Taxonomy of Artifacts in E-D**
  - **Different Artifacts need Different Processing**
  - **Current Work by P-M Jodoin: Case around  $\frac{1}{2}$**
- **Better Cost Function (Automatic Quality Evaluation)**
  - **Pathetic Cases (Blue Noise Criterion Does Not Work)**
  - **Detection of Local Structures (Wavelets, Gabor Functions etc.)**



## Future Work

- **Taxonomy of Artifacts in E-D**
  - Different Artifacts need Different Processing
  - Current Work by Pierre-Marc Jodoin: Case around  $\frac{1}{2}$
- **Be** (Automatic (on))
  - **o** (oise ())
  - **o** (structure (ctions etc.))
- **Sy** (ram (nality)
  - **o** (Proc

al Coherence with E-D

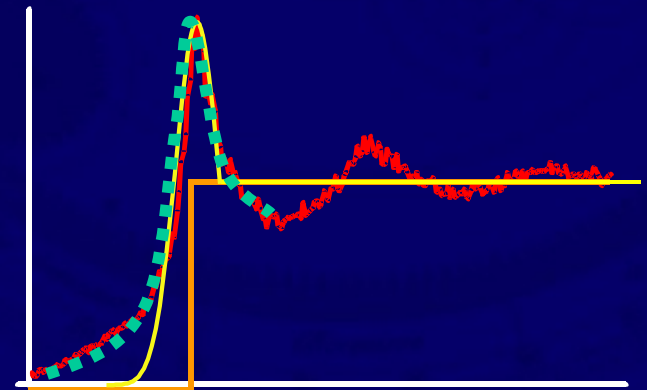
**x2  
Enlargement**

**x2  
Enlargement**



## Future Work

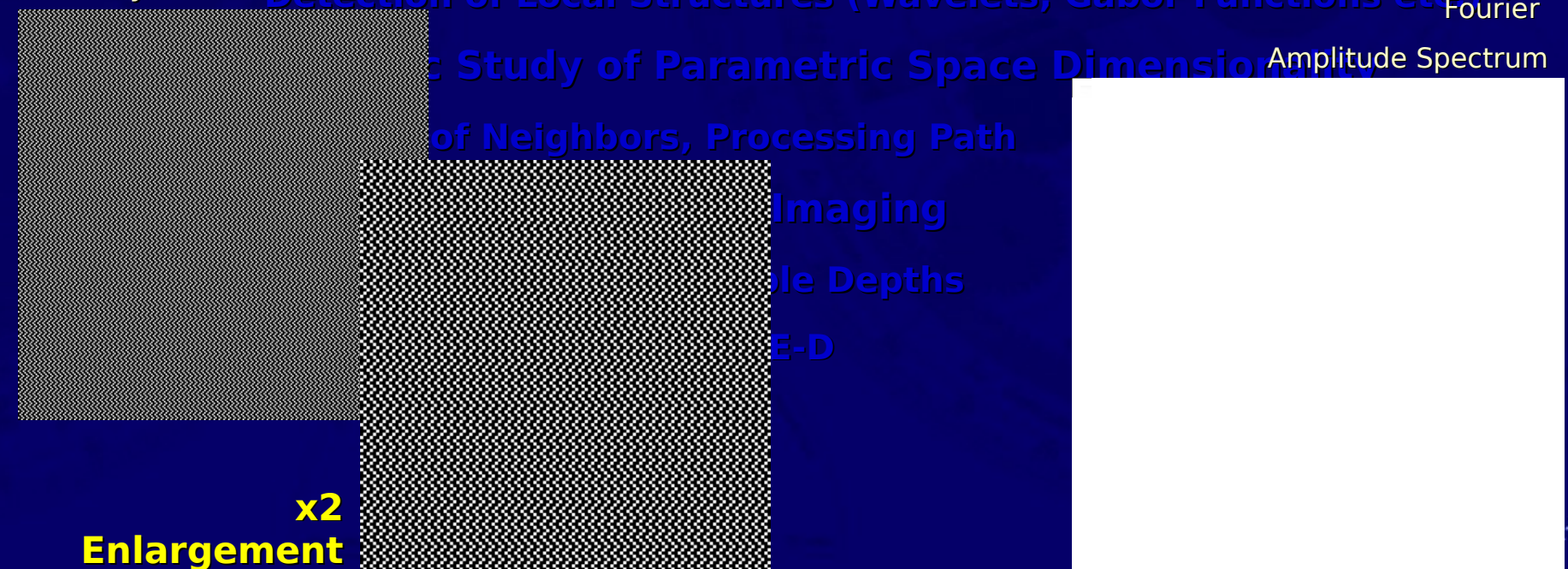
- **Taxonomy of Artifacts in E-D**
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  - Number of Neighbors, Processing Path
- **E-D for Network-Oriented Imaging**
  - Coherence Through Multiple Depths
  - Temporal Coherence with E-D



## Future Work

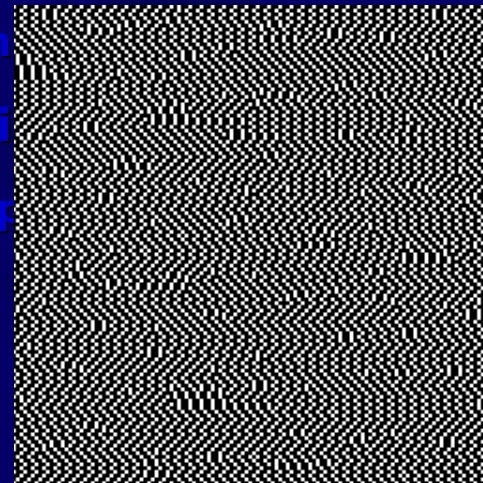
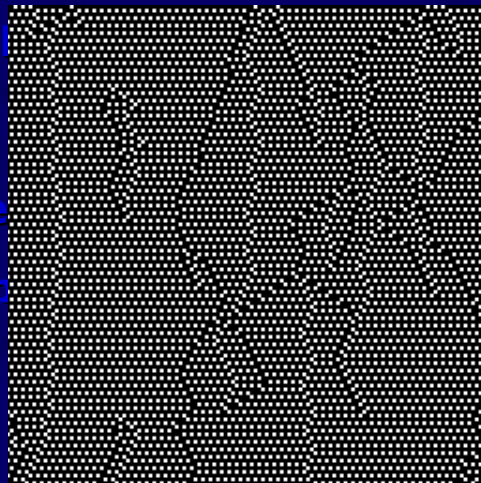
- **Taxonomy of Artifacts in E-D**
  - Different Artifacts need Different Processing
  - Current Work by P-M Jodoin: Case around  $\frac{1}{2}$
- **Better Cost Function (Automatic Quality Evaluation)**
  - **Pathetic Cases (Blue Noise Criterion Does Not Work)**

Intensity level =  $\frac{1}{3}$



## Future Work

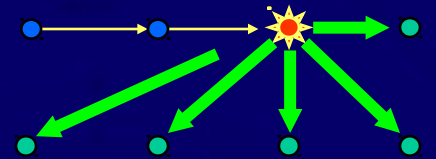
- **Taxonomy of Artifacts in E-D**
  - Different Artifacts need Different Processing
  - Current Work by P-M Jodoin: Case around  $\frac{1}{2}$
- **Better Cost Function (Automatic Quality Evaluation)**
  - Pathetic Cases (Blue Noise Criterion Does Not Work)
  - **Detection of Local Structures (Wavelets, Gabor Functions etc.)**
- **Systematic Study of Parametric Space Dimensionality**
  - Number of Parameters in Processing
- **E-D for Image Compression**
  - Coherence and Image Dependent
  - Temporal Coherence in E-D



## Future Work

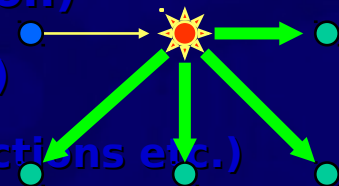
- **Taxonomy of Artifacts in E-D**

- Different Artifacts need Different Processing
- Current Work by P-M Jodoin: Case around  $\frac{1}{2}$



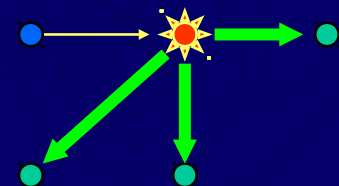
- **Better Cost Function (Automatic Quality Evaluation)**

- Pathetic Cases (Blue Noise Criterion Does Not Work)
- Detection of Local Structures (Wavelets, Gabor Functions etc.)



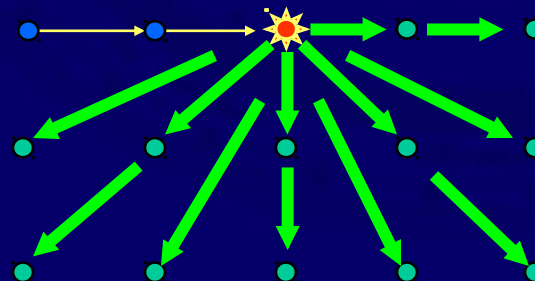
- **Systematic Study of Parametric Space Dimensionality**

- Number of Neighbors, Processing Path



- **E-D for Network-Oriented Imaging**

- Coherence Through Multiple Depths
- Temporal Coherence with E-D



## Future Work

- **Taxonomy of Artifacts in E-D**
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  - Number of Neighbors, Processing Path
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  - Coherence Through Multiple Depths
  - Temporal Coherence with E-D



## Thanks

- Bob Ulichney, Jan Allebach, Gabriel Marcu, Reiner Eschbach, Luiz Velho, and Kevin Parker for insightful discussions and for providing sample images of their techniques
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**Thank You**